

1 Residual canopy cover provides buffering of near-surface temperatures, but benefits are limited under
2 extreme conditions

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9 ABSTRACT

10 Increasing summer temperatures and higher probabilities of extreme heat events have led to
11 concerns about tree damage and mortality. However, insufficient attention has been given to conditions
12 leading to heat-related regeneration failures in temperate forests. To address this managers need to
13 understand how microclimate varies under a range of overstory conditions. We measured air
14 temperatures at 2cm above-ground underneath a gradient of canopy cover on south-facing slopes in
15 recently thinned Douglas-fir stands in western Oregon, USA. To expand the ecological relevance of these
16 data to impacts on regeneration, we created the stress-degree hours (SDH) metric, representing the
17 amount of time - and by how much – temperatures exceeded biologically relevant stress thresholds.
18 Overall, for every 10% increase in canopy cover, maximum temperatures at 2cm were 1.3oC lower, the
19 odds of temperatures exceeding stress thresholds for conifer regeneration declined by a multiplicative
20 factor of 0.26, and the total of SDH decreased by 40%. These reductions are large enough to be worthy
21 of attention when managing for tree regeneration. However, data collected during the Pacific Northwest
22 Heat Dome in June 2021 indicate that with various climate change scenarios and heatwave occurrences,
23 temperatures will be unfavorable for regeneration regardless of overstory cover.

24 2. KEYWORDS

25 Climate change, heatwave, microclimate, forest management, regeneration, Douglas fir, retention
26 harvest, Western Oregon, Silviculture, thinning

27 3. INTRODUCTION

28 Trends of higher summer air temperatures have led to increasing concerns about a loss of tree
29 vigor and mortality (Adams et al. 2017; Hammond et al. 2022). Over the last decade it became evident
30 that much of the tree mortality in selected regions was due to a combination of drought and high
31 temperatures (Yi et al. 2022), whereby the temperatures lead to higher evaporative demand (Grossiord

32 et al. 2020) and additional direct heat damage to cellular processes (Geange et al. 2021). Furthermore,
33 higher probabilities of extreme events (Puettmann 2021; Hammond et al. 2022), such as the recent
34 extreme heat event (“Heat Dome”) in the US Pacific Northwest and British Columbia in 2021, led to
35 additional concerns of widespread leaf damage when temperatures exceeded critical thresholds (Still et
36 al. 2023; Doughty et al. 2023). The alarming effects of hotter average conditions as well as extreme heat
37 events on forest health have already been documented for a range of forest types (Adams et al. 2017;
38 Hammond et al. 2022). Given future climate predictions, scientists and managers continue to explore
39 opportunities to increase individual tree and forest level resilience to a hotter and more extreme
40 climate.

41 Silviculture has a long history of using density management approaches to maintain and
42 increase tree vigor as well as protect trees and stands from pests, pathogens, and damage from weather
43 events such as strong winds (Chmura et al. 2011; Park et al. 2014). As global change emerged as a major
44 threat to forests, these same principles have been used for climate change adaptation. In the past,
45 thinning has primarily been used to improve growth of residual trees. Recently, the focus has shifted
46 from solely promoting growth to also increasing resilience to drought and hotter temperatures (Halofsky
47 et al. 2016; Bottero et al. 2017). Much of this shift has focused on mature trees (Sohn et al. 2016) with
48 less attention on methods to protect or promote tree regeneration (Walck et al. 2011) and understory
49 development (Christiansen et al. 2022) under climate change, even though regeneration failures have
50 increasingly become an issue of interest for foresters (Dey et al. 2019), especially on drier sites (Dodson
51 and Root 2013; Boucher et al. 2020) and whole regions (Petrie et al. 2023; Crockett and Hurteau 2024).
52 There has also been growing interest in natural regeneration due to its inherent variability and potential
53 to aid in the development of heterogenous stand conditions associated with late successional forests
54 (Donato et al. 2012). Because of the different sensitivities between seedlings and larger trees (Rollinson
55 et al. 2021), a better understanding of the ability of silvicultural actions to mitigate the effects of climate

56 change on regeneration is needed to address concerns of regeneration failures (Dey et al. 2019;
57 Rollinson et al. 2021) and meet multiple objectives including creating late successional forest conditions.

58 The general impact of topography (Scherrer and Körner 2011; Meineri et al. 2015) and forest
59 canopy cover (De Frenne et al. 2019) on microclimate conditions in the understory is well documented
60 in the context of microrefugia and their impacts under normal and extreme climate conditions
61 (Finocchiaro et al. 2024). Forest management practices have capitalized on the buffering capacity of
62 stand structure for decades by implementing a variety of silvicultural practices. For example,
63 shelterwood and uneven-aged silvicultural systems have long been used to modify microclimate and
64 provide suitable conditions for seedling growth and survival through retention of overstory trees
65 (Ashton and Kelty 2018; Palik et al. 2020), both in areas with potentially damaging summer
66 temperatures (Childs et al. 1985); but see Valigura and Messina (1994) and areas where frost is likely to
67 damage tree seedlings (Granberg et al. 1993; Holgén and Hånell 2000; Langvall and Ottosson Löfvenius
68 2002). The physical drivers of the relationship between forest canopy cover and microclimate conditions
69 in the understory are well understood (Geiger et al. 1995; Campbell and Norman 1998), including
70 modifying processes such as interception and attenuation of incident solar radiation, air mixing,
71 precipitation interception and throughfall, windspeed, and humidity (Geiger et al. 1995; Kovács et al.
72 2017). Recently, there has been growing interest in gaining a deeper understanding on the effects of
73 different forest management actions on microclimate conditions and the implications for regeneration
74 and understory composition (De Frenne et al. 2013).

75 Studies have confirmed the importance of the fine scale-variation in microclimate on seedling
76 establishment and tree recruitment, and in the context of climate change this issue is gaining more
77 attention (Halpern et al. 2012; Peck et al. 2012; Swanson et al. 2023). Larger regional variation in
78 atmospheric processes are also a main driver of microclimate conditions and can affect the relative
79 influence of canopy cover (Finocchiaro et al. 2024). Extreme heat events such as the Pacific Northwest

80 June 2021 Heat Dome may overwhelm the ability of canopy cover to reduce heat stress in the
81 understory. Areas with increases in wildfire smoke in the atmosphere may also impact this relationship
82 by absorbing and scattering incoming solar radiation at global (Tosca et al. 2013), regional, and local
83 scales (Price et al. 2016) as well as impacting stream temperatures (David et al. 2018). To facilitate forest
84 management in a variety of settings in a rapidly warming world, we need to understand how
85 microclimate conditions in forest understories change across a gradient of partial overstory cover,
86 whether the specific location of the overstory canopy in relation to solar angle affects the microclimate,
87 how regional conditions affect microclimate relative to canopy cover, and how the conditions relate to
88 relevant ecological processes, such as regeneration. To gain such understanding, we set up a study with
89 the following objectives:

- 90 1. Determine the effect of varying amounts and spatial arrangements of canopy cover on summer
91 maximum near-surface temperatures.
- 92 2. Use previously established heat stress responses of seedlings from laboratory studies to assess
93 potential temperature-induced stress to conifer seedlings and germinating seeds under varying
94 overstory canopy cover conditions.
- 95 3. Describe how future climate conditions, including heat waves and wildfire smoke, may affect the
96 influence of canopy cover on understory temperatures.

97 By sampling across a gradient of canopy cover our findings can be used to inform decisions
98 regarding a variety of density management practices, including treatments that result in higher spatial
99 variability in residual tree density (Puettmann et al. 2009; Palik et al. 2020). Thus, our findings are
100 relevant for a wide variety of conditions including homogenous thinning prescriptions in even-aged
101 stands, as well as variable density treatments designed to achieve a variety of objectives (Puettmann et
102 al. 2016; Franklin and Donato 2020).

103 4. MATERIALS & METHODS

104 4.1 STUDY AREA

105 This study was conducted in 11 stands (< 15km apart) in the Big Blue Project area in the Upper
106 Blue River Watershed on the Willamette National Forest in western Oregon (Fig. 1). The Big Blue Project
107 area was selected due to the similarity in conditions (slope, aspect, stand histories and conditions), the
108 proximity to the meteorological stations at the H.J. Andrews Experimental Forest, availability of recent
109 LIDAR data (2020), and because several stands in this area were recently commercially thinned (< 5
110 years prior) which resulted in a variety of overstory densities and canopy covers and minimal understory
111 vegetation. The U.S. Forest Service plant association for this area is *Pseudotsuga menziesii*/Acer
112 *circinatum*- *Berberis nervosa* (Douglas-fir/vine maple - Oregon grape) (Dyrness et al. 1974).

113 All study sites are covered by approximately 50-year-old even-aged monoculture Douglas-fir
114 plantations that were recently thinned. Within stands the post-thinning tree spacing was targeted to be
115 homogenous. However, variability in microsites, past conditions, and initial spatial arrangement of trees
116 growing in operational settings resulted variable spatial arrangements at smaller scales. Among stands
117 the prescribed minimum spacing between trees post-thinning varied from 4 to 6 meters (USDA Forest
118 Service 2009 (Table S4). Similar prescriptions in these types of stands have shown to result in
119 establishment of more and vigorous understory vegetation, including tree regeneration (Beggs et al.
120 2005; Kuehne and Puettmann 2008; Puettmann et al. 2016). For more detail on stand conditions and
121 thinning prescription see supplemental materials.

122 The study sites ranged in elevation from 630m to 1086m. The mean monthly temperatures for
123 this area ranged from 2.6 °C in January to 19.3 °C in July and annual precipitation averages 2.17m based
124 on 30-year normal data from 1991-2020 (PRISM Climate Group, 2022). This area is also characterized by
125 highly seasonal precipitation and a mostly dry growing season. During this study, conducted from June

126 29, 2021-September 25, 2021, the mean daily near-surface air temperature ranged from 11.3 °C to 22.2
127 °C at the Primary Meteorological Station (PRIMET) of the H.J. Andrews Experimental Forest (located at
128 430m elevation approximately 6.5 km south of the study sites) (Daly and McKee n.d.). The study area
129 experienced an unprecedented heat wave (“Heat Dome”) between June 25 and July 3rd, 2021, with
130 maximum air temperatures at 1.5m reaching 46 °C, as well as several subsequent smaller heat waves
131 with maximum air temperatures ranging from 38- 40.5 °C. During the Heat Dome event, maximum near
132 surface temperatures at our study sites was 44.1 °C on average and the absolute maximum temperature
133 recorded was 57.4 °C. In upper tree canopies at the Andrews Forest, foliar temperatures exceeded 50 °C
134 and stayed above 40 °C for at least 26 hours (Still et al. 2023).

135 The study area was also heavily impacted by smoke in early and mid-August of 2021 due to the
136 Middle Fork Complex and Washington Ponds Fires. While these events complicate the interpretation of
137 our data, such conditions may be indicative of future climate conditions of this area given the predicted
138 increase in duration and severity of heatwaves (Mazdiyasi and AghaKouchak 2015) and large wildfires
139 (Abatzoglou 2013). This event allowed us to quantify the impacts of wildfire smoke on near-surface air
140 temperatures as an illustrative example for the future.

141 4.2 STUDY DESIGN

142 To capture the microclimate conditions at locations relevant to germinating seeds and young
143 seedlings we used Tomst TMS-4 temperature sensors (Wild et al. 2019), which capture climate and soil
144 moisture conditions near ground level. These sensors also had a small radiation shield installed (Wild et
145 al. 2019). For this study we only used the air temperature measured at 2cm above ground, since this
146 height is heavily influenced by the soil surface temperature and reasonably represents the climate
147 experienced by newly germinated and very young seedlings, which are the life stages most susceptible
148 to heat damage (Bell et al. 2014).

149 We installed TMS-4 sensors at 20 locations (1 sensor per location) in 11 different stands with
150 specific locations selected by a stratified random sample and nested design. Sampling was stratified by
151 levels of overstory canopy cover. The range of each level was 10% canopy cover, resulting in 7 levels
152 from 15% to 85%. There were three sensors per level apart from the 46-55% level containing four
153 sensors and the 75-85% level with only one sensor. This was done to ensure the full range of canopy
154 cover was sampled while accounting for the larger variability in conditions at the mid-levels compared to
155 higher canopy cover (Table S1). Slope, aspect, and elevation also affect the potential direct incident
156 radiation and therefore the microclimate conditions, but we did not have sufficient resources to cover
157 all those gradient combinations and thus focused on a restricted set of conditions. Sites were nested
158 within stands and limited to slopes of less than 25 degrees, with a southern aspect (S or SW), and
159 between 600-1100m of elevation. South- or southwest-facing slopes were chosen because they receive
160 the highest amount of incident radiation in the Northern Hemisphere and are therefore more likely to
161 be warmest and have the most limiting microclimates for tree regeneration. Mid-slope positions were
162 chosen to limit the effect of cold-air drainages or hill-shade. See table S1 for additional site-specific
163 information.

164 One hundred potential sensor locations within areas with the target topographic conditions
165 were randomly generated and stratified by canopy cover within a GIS. Each of the potential locations
166 were randomly assigned a priority level of 1-5 such that each priority level contained the full range of
167 canopy cover conditions. Levels 2-5 were back-up locations and were only used if the level 1 site was
168 determined to be unsuitable in the field due to excessive slash, microtopography conditions such as
169 seeps, or other factors that might have affected the near surface temperature other than canopy cover
170 and therefore might have biased the results. To limit spatial auto-correlation in relation to microclimate
171 variables the minimum distance between sites was 100m (Chen et al. 1995; Baker et al. 2016). The
172 stratified sampling design resulted in 20 sensor locations in 11 different stands. Apart from one stand

173 (Big B 660), which had four, there were only one or two sensors per stand, as the focus of this study was
174 to examine microclimate conditions across a gradient of overstory cover and not within-stand
175 microclimate variability.

176 4.3 TEMPERATURE DATA PROCESSING

177 Our primary interest was to understand how canopy cover influences air temperatures in
178 understory conditions in the context of tree regeneration. Specifically, we were interested in differences
179 in daily maximum air temperatures and relating those conditions to potential seedling/germinating seed
180 stress. We summarized the data to weekly averages of daily maxima air temperature ($^{\circ}\text{C}$). To estimate
181 the impact of heat on seedling stress we used a novel application of the degree-day concept: stress
182 degree hours (SDH). Instead of calculating the accumulation of heating units above a minimum
183 threshold that reflects physiological processes and the initiation of plant growth, we used a high
184 temperature threshold above which photosynthetic damage typically occurs in tree seedlings, and
185 calculated the accumulation of heating units above this threshold over time (Baskerville and Emin 1969).
186 The base temperature in these calculations (40°C) was derived from photosynthetic responses of
187 Douglas-fir seedlings exposed to simulated heatwaves of temperatures from 25 to 61°C (Marias et al.
188 2017). We calculated weekly averages of daily accumulated stress degree hours by subtracting 40°C
189 from hourly average temperatures. Negative values were reassigned to 0 and the positive values were
190 summed to a daily accumulation. The daily accumulation of SDH was then averaged for each week (Cook
191 et al. 2024). The last two weeks of the study (Sep 14 - 28) were removed since temperatures were
192 significantly cooler due to changing seasons and seedlings were not stressed based on our definition.
193 The resulting metric used in our analysis was the weekly average of daily accumulated SDH. It represents
194 a combination of the number of hours over 40°C , as well as how much higher than 40°C the hourly
195 average temperature was during these periods.

196 4.4 CALCULATING EXPLANATORY VARIABLES

197 To investigate our primary objectives, canopy cover was measured using two different
198 approaches. First, we used canopy closure values from convex spherical densiometer measurements,
199 which integrate across the sky ignoring the azimuth to represent total canopy cover (referred to as 360°
200 measurement in tables and figures). This was selected as it was a simple method that foresters can
201 apply easily in the field. Second, to examine the effect of direct shading and sun flecks due to canopy
202 gaps and the orientation of canopy cover relative to the sun position we used LiDAR data from a June
203 2020 flight to isolate the canopy cover that shades the sensor based on time of day, solar angle and
204 azimuth, aspect, slope, and average tree height (Fig. S1). For this, we quantified canopy cover that
205 provided shade during three time periods, from 9am-12pm, 12pm-3pm, and from 9am-3pm. This
206 resulted in four canopy cover variables: one 360° field measurement using a densiometer and the three
207 LiDAR derived metrics (Table 1).

208 Additional explanatory variables that varied by sensor location included heat load and elevation.
209 Heat load was calculated following McCune (2007) using latitude, slope, and aspect of each sensor
210 location to represent the potential incoming solar radiation at each sensor in the absence of canopy
211 cover and assuming clear skies. Elevation was determined using the digital elevation model (DEM) from
212 LiDAR. The weekly average of the daily maximum air temperature from Central Meteorological Station
213 (CENMET) at the H.J. Andrews Experimental Forest was included in the models to account for regional
214 climate conditions and autocorrelation among weeks (Daly and McKee n.d.).

215 In the initial model fitting temporal autocorrelation was not accounted for by using a correlation
216 structure, indicating the variables in the models did not sufficiently explain the trend through time. We
217 hypothesized that this was due to the timing of nearby wildfires, as the study area was impacted by
218 smoke throughout most of August. Smoke and high particulate concentrations in the air increases

219 scattering of incoming solar radiation and decreases the amount of direct radiation on the sensor
220 (Rastogi et al. 2022). Since the TMS-4 only had a small radiation shield this change in conditions likely
221 affected the temperature recorded by the sensor, especially the maximum daily temperature.
222 Additionally, temperature at 2cm is heavily influenced by the radiative heating of the soil surface and
223 was likely impacted by changes in direct radiation (Campbell and Norman 1998). Consequently, we
224 included the weekly average daily accumulation of incoming short-wave radiation (J/m²/day) measured
225 at CENMET to account for the reduction in short-wave radiation due to wildfire smoke and subsequently
226 the temporal autocorrelation. Although heat load and incoming short-wave radiation represent the
227 amount of solar radiation received, both variables were included in the analysis since heat load
228 accounted for site differences while measured incoming short-wave radiation accounted for differences
229 through time and the impact of smoke. See tables S1 and S2 for additional information on how the
230 variables varied by sensor location and week.

231 4.5 STATISTICAL ANALYSIS

232 To develop an understanding of the relationship between canopy cover and maximum near-
233 surface temperatures, we used weekly average daily maximum temperature (°C) at 2cm above the
234 ground as the response variable in four linear mixed models. These models used canopy cover, CENMET
235 weather station 1.5m screen height air temperature, elevation, heat load, and incoming shortwave
236 radiation as continuous explanatory variables, with sensor location nested in harvest unit as a random
237 effect. Each of the four models used a different canopy cover measurement described above and in
238 Table 1. To avoid overfitting due to the relatively small number of sites, we limited the number of
239 explanatory variables in the models. To account for this and still test if average canopy cover from the
240 densiometer or the LiDAR-derived canopy cover measurements better explain near-surface air
241 temperature, we used a model comparison approach.

242 All linear mixed models were extended to allow for among-week correlations of the errors
243 within sensor locations nested in harvest unit using an auto-regressive correlation of lag 1 (AR1). This
244 correlation structure estimated a single correlation used to describe how errors within weeks become
245 less similar with increasing time between measurements. Assumptions of constant variance and
246 normality of errors for each model were assessed visually using plots of the normalized residuals (Fig.
247 S2). No problems were noted. Sensor locations were assumed to be independent of each other due to
248 the minimum inter-sampling distance of 100m (Chen et al. 1995; Baker et al. 2016). Harvest unit was
249 included in the models as a random effect to account for any with-in unit spatial autocorrelation
250 (distance between sensors in different harvest units was > 1km). Delta Akaike's corrected information
251 criteria values ($\Delta AICc$) were used to compare evidence for model support following Burnham et al.
252 (2011) where a $\Delta AICc$ value less than 7 indicates no difference in model support. Pseudo R^2 were
253 calculated following Efron (1978) and regression coefficients were also used to interpret model fit and
254 the relationship between each response variable and the four canopy cover measurements (Fig. 2).

255 To assess the likelihood and amount of temperature-induced stress that seedlings and
256 germinating seeds may experience, we used a hurdle model approach. This was necessary given that
257 SDH calculations resulted in positive continuous data with a point mass at zero, which make fitting a
258 single statistical model difficult and prone to bias (Brooks et al. 2017). The hurdle model was comprised
259 of two models: one to examine whether the threshold for SDH accumulation (> 40 °C for one hour) was
260 passed and another to examine the amount of SDH accumulation (positive values only) (Table 1). For the
261 first model, the SDH weekly average daily accumulation was transformed into a binary variable with
262 values of 1 for weeks above 0.25 SDH and 0 for weeks below 0.25 SDH. These small values of SDH
263 (<0.25) negatively affected the interpretability of the second component of the hurdle model. Given the
264 accuracy level of the sensor (0.5°C) (Wild et al. 2019) we determined values less than 0.25 were not
265 different than zero. From a plant physiological perspective these small values are negligible (Marias et

266 al. 2017), thus setting the values to zero allowed for increased interpretability of the results while
267 keeping the variables tied to plant physiological principles.

268 This binary variable of presence or absence of SDH, which represented whether a biologically
269 relevant amount of stress ($SDH > 0.25$) accumulated, was used as the response variable in four binomial
270 generalized linear mixed models using a logit link (also referred to as logistic regression). This type of
271 model was chosen because of the ability to estimate the probability of an event (SDH accumulation)
272 occurring and because binomial distributions were appropriate for presence/absence data. Standardized
273 continuous explanatory variables of canopy cover, open air temperature, elevation, heat load, and
274 incoming shortwave radiation and a random effect of sensor nested in harvest unit were also included in
275 the models (Table 1). The continuous fixed effects were standardized by subtracting the mean from each
276 value and dividing by the standard deviation to accommodate for the difference in scales between
277 variables. Plots of simulated residuals relative to fitted values were examined and no unusual patterns
278 or overdispersion were noted (Fig. S2).

279 In the second component of the hurdle model the weekly average daily accumulation of SDH for
280 weeks when the average was greater than 0.25 was log transformed and used as a response variable in
281 a family of four linear mixed models with the same fixed and random effects as previous models (Table
282 1). Based on a graphical assessment of residual plots, a natural logarithm transformation of weekly
283 average daily accumulation of SDH adequately stabilized the variance and residuals were sufficiently
284 symmetric and approximately normal (Fig. S2). The delta corrected Akaike's information criteria value
285 ($\Delta AICc$), Pseudo R^2 , and coefficients were used to interpret model support and fit and the relationships
286 between each response variable and the four canopy cover measurements. Pseudo R^2 was calculated
287 following Nakagawa and Schielzeth (2013). The nlme package was used to fit the linear mixed models
288 (Pinheiro et al. 2022) and the lme4 package was used to fit the binomial generalized linear mixed models
289 (Bates et al. 2015). Analyses were done with R version 4.1.2 (2021).

290 To portray how these relationships are biologically relevant under varying regional climate
291 scenarios (Objective 3), the models containing canopy cover measured with a densiometer were used to
292 describe the relationship between canopy cover and the three response variables (weekly average daily
293 maximum temperature, presence of SDH, and accumulation of SDH) under four different climate
294 scenarios: 30-year normal, 3 °C of warming added to the normals, the daytime average conditions during
295 the June 2021 Heat Dome, and the average air temperature during the hottest day of the Heat Dome
296 (Fig. 3). For all scenarios a prediction dataset was created using the models developed in objectives 1
297 and 2 (See Supplement for more details). These 24 prediction datasets (3 response variables, 4 climate
298 change scenarios, 2 smoke conditions) were then used to predict maximum temperature, probability of
299 SDH accumulation, and the amount of SDH accumulation across the range of canopy cover for each
300 scenario.

301 5. RESULTS

302 The study results demonstrate that –after accounting for the influences of topography - canopy
303 cover reduces near surface temperatures. We found a reduction in maximum (and mean temperatures)
304 in the understory with increasing canopy cover (Fig. 2). The model comparison results suggest that
305 accounting for the specific location of trees that provide shade did not improve the ability to predict the
306 impact of canopy cover in reducing near-ground temperatures (Table 2). Low $\Delta AICc$ values (< 7) for
307 models used to address objective 1 suggest that there is no difference in support among all models
308 (Burnham et al. 2011) suggesting that the method used to collect canopy cover (i.e., LiDAR versus
309 densiometer-derived data) did not influence model support. However, the model containing
310 densiometer canopy cover better predicts the mean weekly average daily maximum air temperature at
311 2cm, as indicated by a slightly higher pseudo R^2 (0.82) (Table 2). Thus, the selected model used weekly
312 average daily maximum air temperature as a response with fixed effects of densiometer-derived canopy

313 cover, open air temperature at 1.5m, elevation, heat load, and weekly average daily accumulation of
314 incoming shortwave radiation and sensor location nested in harvest unit as a random effect.

315 Similarly, the results indicate that any stress, as quantified by SDH, that seedlings may
316 experience due to heat is not influenced by the orientation of cover relative to solar position of the
317 shading trees. The $\Delta AICc$ (7.4) for the first component of the hurdle model suggested there was no
318 difference in support for any individual model in estimating the probability of SDH accumulation (Table
319 2). As with the analysis of temperature reductions due to overstory canopy cover, utilizing data collected
320 with a densiometer for assessment of stress appears to be just as valid as using LiDAR data which is
321 much more challenging to collect and use. The pseudo R^2 for the model with densiometer-based
322 measurement data was slightly higher (0.72) (Table 2). In regard to the amount of SDH accumulation,
323 the second component of the hurdle model, the model containing the densiometer measurement was
324 better supported by the data than the LiDAR-based measurements ($\Delta AICc$ 14.48). The difference in
325 pseudo R^2 between models is also larger for this comparison (Table 2).

326 Based on these results, we quantified the impact of residual trees in terms of mediating high
327 temperature conditions using the models containing the densiometer (360°) measurement. After
328 accounting for elevation, heat load, incoming shortwave radiation, and regional, i.e., weather station
329 screen air temperature at 1.5m, every 10% increase in canopy cover (measured using a densiometer)
330 was predicted to decrease the mean weekly average daily maximum at 2cm by 1.3 °C (95% CI 0.4 to
331 2.2 °C), the odds of accumulating SDH by a factor 0.26 (95% CI 0.06 to 0.59), and the median weekly
332 average daily accumulation of SDH by 40% (95% CI 20-55%) (Table 3; Fig. 2). These results are based on
333 the range of temperatures observed during the summer of 2021 which, as mentioned above, was hotter
334 than normal. See table S5 for estimates for elevation and heat load index.

335 The results can also be used to quantify how higher canopy cover led to lower potential for heat
336 stress in vegetation near the ground under current conditions. The relationship between the probability
337 of accumulating stress and canopy cover indicated that on south-facing slopes, maintaining at least 60%
338 canopy cover under normal temperature regimes may avoid temperature stress for seedlings. In stands
339 with higher canopy cover plant stress in the understory is less likely (probability of stress) and less
340 intense (accumulation of stress hours). Distinguishing the probability and absolute amount of stress
341 provides additional insights. For example, in stands with 40% canopy cover the probability of SDH is
342 much higher than at 60% cover but the median average daily accumulation only 4 SDH (Fig. 3). The high
343 probability, but low amounts of SDH in such stands indicates that the temperature buffering as
344 experienced by seedlings may be quite substantial in terms of reducing heat stress.

345 Simulations of potential future climates indicated that under 3°C of warming of average
346 summertime temperatures and during extreme events, such as the 2021 Heat Dome, the buffering
347 effect of canopy cover was not strong enough to prevent temperatures at 2cm from crossing the
348 biologically relevant 40°C threshold even at high canopy cover (Fig. 3). However, the presence of wildfire
349 smoke during the study period resulted in lower near-surface temperatures, suggesting additional
350 buffering effects. The smoke impact on the near-surface temperature maxima (and means) was similar
351 to the temperature reduction caused by an increase of 15% in canopy cover (Brackett et al. 2022).

352 6. DISCUSSION

353 This study confirmed the large role of overstory trees in influencing the understory temperature
354 regime in stands without much understory vegetation (Rambo and North 2009). It demonstrated that
355 greater canopy cover resulted in reductions of maximum temperatures and heat-related stress levels
356 that were sufficiently large to be ecologically relevant to tree germinating seeds and other understory
357 plants (Jansen et al. 2014; Marias et al. 2017). However, under climate change and heatwave scenarios,

358 conditions on south-facing slopes in our study region will likely be unfavorable for regeneration. The
359 spatial variability in conditions (Macek et al. 2019) as well as variability in species and individual
360 responses to heat and moisture stress (Marias et al. 2017; Guha et al. 2018) indicate that while growth
361 and survival rates may decline, large-scale regeneration failure is still unlikely in this region (De
362 Lombaerde et al. 2022). Forest managers whose goal is to create structural diversity through lower
363 stand densities and understory regeneration will need to develop decision and risk analysis tools that
364 can incorporate the buffering capacity of higher tree cover and topography to create sufficient spatial
365 variability in microclimate conditions for regeneration success to occur at the stand and landscape level.
366 We also found that the presence of wildfire smoke reduced near-surface temperatures. The reduction in
367 radiative heating as particles in wildfire smoke reflect and absorb incoming solar radiation (Stone et al.
368 2011), apparently reduced temperature near the soil surface.

369 Our study documented the buffering effect of canopy cover in regard to high temperatures and
370 specific biological relevance by sampling near the soil surface, where seedlings and germinating seeds
371 are most sensitive to heat (Harper 1977; Rollinson et al. 2021). The low sensor height of 2cm above
372 ground made direct comparison of our results with those of other studies difficult. At first glance, the
373 1.3 °C decrease in maximum temperatures for every 10% increase in canopy cover found in this study is
374 larger than effects found in most other studies. For example, in western Washington, conditions at 1
375 meter above ground show only a reduction of approximately 4 °C under an unthinned control (90%
376 canopy cover) and ~50% canopy cover (Heithecker and Halpern 2006). The current results suggest a
377 difference of 5.2 °C in near-surface temperatures for the same canopy cover difference. This discrepancy
378 was partially due to the differences in sampling height (2cm vs 1-2m), as large air temperature gradients
379 commonly occur in the first few meters above the soil surface, and our sampling was limited to south-
380 facing aspects only (Geiger et al. 1995). Other studies found maximum air temperatures to be lower
381 under closed canopy versus stands thinned to various degrees or canopy gaps anywhere from 0.6 °C to 5

382 °C, measured between 1 and 5m above ground (Heithecker and Halpern 2006; Kovács et al. 2020).
383 Additionally, studies that examined the effect of thinning and shelterwood treatment on soil
384 temperatures at 20 mm below ground found a larger difference of ~7 °C between controls and
385 treatments (Childs et al. 1985; Peck et al. 2012). Our results provide support to these findings, as they
386 fell in between those results, confirming that maximum air temperatures near the surface are expected
387 to be lower than soil surface temperatures but higher than air temperatures further from the ground
388 due to the conductive heating from the soil surface (Geiger et al. 1995; Campbell and Norman 1998).

389 Much of the research that quantifies seedling survival and performance following retention
390 harvests has focused on availability of light as a limiting factor (Gagnon et al. 2003; Powers et al. 2008;
391 Peck et al. 2012), although selected research on shelterwood systems has examined soil temperature
392 effects on seedling survival and growth (Childs et al. 1985; Man and Lieffers 1999) and many have
393 looked at climate conditions alone (Granberg et al. 1993; Valigura and Messina 1994; Langvall and
394 Ottosson Löfvenius 2002). Childs and co-authors (1985) found that on south or west aspects in
395 southwest Oregon where temperatures are generally higher than our study sites, shade from
396 shelterwoods was beneficial in protecting seedlings from heat damage. While we did not directly
397 examine regeneration rates or seedling survival or performance, the interpretation of the temperatures
398 near the soil surface and SDH was based on physiological studies (Marias et al. 2017; Rank et al. 2022)
399 and thus reflected their ecological relevance, especially for germinating seeds and seedlings and other
400 understory plants. The relationships between canopy cover and SDH in our study confirm results from
401 studies that examined the probability of regeneration after varying fire severity and climate scenarios
402 (Willms et al. 2017; Davis et al. 2023). The associated lower post-fire canopy cover will result in limited
403 temperature buffering and associated higher temperature stresses of the regenerating vegetation. For
404 example, Davis et al. (2023) showed the benefits of retained canopy cover on natural tree regeneration
405 was likely due to a combination of shading and seed availability. Retained canopy cover may provide

406 additional benefits to natural tree regeneration through reducing competition from understory
407 vegetation (Devine and Harrington 2008; Dodson et al. 2014).

408 The emphasis on light availability as a limiting factor on seedling survival and performance
409 following retention harvests highlights a key tradeoff of leaving higher canopy cover to buffer
410 understory temperatures (Peck et al. 2012; Käber et al. 2023). Many commercially valuable species are
411 shade intolerant, and higher overstory canopy cover can negatively impact regeneration through
412 reduction of light availability, potentially outweighing benefits of reduced microclimate temperatures
413 (Gray and Spies 1997; Brandeis et al. 2001; Ashton and Kelty 2018). This trade-off can be addressed
414 through incorporating topographical conditions (Scherrer and Körner 2011; Meineri et al. 2015) and
415 diversification of management goals. If high temperatures are the main concern for understory
416 establishment following a harvest, foresters should target north-facing or other topographically buffered
417 areas (Carnicer et al. 2021), leave higher canopy cover initially to provide shading during initial stages of
418 regeneration that are most temperature limited followed by a second entry to improve light availability
419 (Devine and Harrington 2008; Shatford et al. 2009), retain higher canopy cover and shift goals toward
420 establishment of a shade-tolerant cohort, or a combination of these approaches (Kuehne and
421 Puettmann 2008).

422 Additionally, there may also be tradeoffs associated with water availability and competition
423 between residual trees and understory regeneration following retention harvests (Gray et al. 2002;
424 Devine and Harrington 2008). Water availability for seedlings is influenced by competition from
425 overstory trees and other understory vegetation (Devine and Harrington 2007), overstory canopy
426 interception of precipitation, evaporative demand on soil moisture, and the interaction of these effects
427 (Aussenac 2000). The presence of an overstory canopy reduces through fall of precipitation (Geiger et al.
428 1995) but may also reduce soil evaporation rates through shading and litter deposition (Aussenac 2000;
429 Floriancic et al. 2023). In this study we examined the most extreme topographical positions for

430 microclimate conditions where it is likely the benefit of reduced evaporative demand outweighs
431 reduction in throughfall. While overstory trees may negatively affect regeneration through direct
432 competition (Balandier et al. 2006; Devine and Harrington 2008; Riegel et al. 2013), there is evidence
433 that under partial canopies competition from other understory vegetation, which has a larger effect on
434 regeneration, is reduced (Smidt and Puettmann 1998; Montgomery et al. 2010; Dodson et al. 2014).
435 Therefore, the overall competition regeneration experience following retention harvests may be lower
436 than in clear cut settings without vegetation management (Montgomery et al. 2010).

437 While long-term trends of increasing average temperatures have and will continue to impact
438 forest ecosystems, extreme events pose a larger threat to forest health and the provision of ecosystem
439 services (Breshears et al. 2021; Puettmann 2021; Hammond et al. 2022). The June 2021 Northwest Heat
440 Dome shattered temperature records and caused significant foliar damage and mortality throughout the
441 Pacific Northwest (Still et al. 2023) and occurred during our study period. The results of our analysis also
442 indicated that even under high residual canopy cover, temperatures at 2cm are well above thresholds
443 for heat-stress impacts on plants (Jansen et al. 2014). This is of major concern as it indicates that during
444 extreme events understory conditions will likely not be sufficiently protected from damaging and
445 potentially lethal temperatures on south-facing slopes by canopy cover alone. In fact, the Heat Dome
446 conditions led to substantial seedling mortality in the region as well as damage to adult trees due to the
447 heat (Still et al. 2023). Assuming such heat events are becoming more common, managers planning
448 retention harvests with the goal of recruiting a second age cohort through regeneration and/or of
449 establishing vigorous understory vegetation (Puettmann et al. 2016; Franklin and Donato 2020) may
450 need to retain a higher canopy density after harvests in the future especially in topographically
451 vulnerable areas (Meineri et al. 2015; Finocchiaro et al. 2024). However, leaving more canopy may shift
452 the species composition of the regeneration to more shade-tolerant species (Kuehne and Puettmann
453 2008) and will reduce the establishment of light demanding early successional species (Puettmann et al.

454 2016). This may be problematic, as light demanding, early successional tree species are also more likely
455 to be drought tolerant (Niinemets and Valladares 2006).

456 The concerns around extreme events coupled with the reduction in microclimate temperatures
457 due to wildfire smoke highlights that as disturbance regimes and regional climate conditions shift under
458 global change, new dynamics and interactions will arise that demand different approaches (Puettmann
459 2011; Tosca et al. 2013). Despite the increasing presence of wildfire smoke across the western U.S., little
460 is known about how smoke affects plants directly through chemical and physical interactions or
461 indirectly through altering climate and atmospheric patterns from microclimate to global scales (Tosca
462 et al. 2013; Price et al. 2016; McKendry et al. 2019). However, recent studies have documented
463 increases in forest photosynthesis due to diffuse light (Rastogi et al. 2022), but also stomatal occlusion
464 and suppression of gas exchange and photosynthesis (Riches et al. 2024). Wildfire smoke has also been
465 shown to disrupt global air circulation patterns and affect precipitation along the equatorial zone (Tosca
466 et al. 2013).

467 It is important to note that the variability in microclimate temperatures at a landscape scale was
468 not captured in this study, as we limited our sampling to south-facing slopes. However, the general
469 principles of temperature buffering should apply to north-facing slopes, where absolute temperatures
470 and thus the stress vegetation experiences will be lessened (Geiger et al. 1995). Also, since forest
471 structure, including canopy height, layers, and composition influence temperatures (Rambo and North
472 2009), e.g., through air mixing, our findings may not be applicable to old-growth (Wolf et al. 2021) or
473 very young stands (Kovács et al. 2017). However, the stands selected for this study were representative
474 of the age class and structure at which the first commercial thinning is typically conducted on managed
475 forests in the PNW as well as in other regions (Ashton and Kelty 2018; Franklin and Donato 2020). Even
476 if not specifically desired after thinning operations, regeneration of tree seedlings and other vegetation,

477 is a major factor in forest development as the start of understory reinitiation (Oliver and Larson 1996;
478 Kuehne and Puettmann 2008; Dodson et al. 2014) and the associated habitat conditions (Hagar 2007).

479 For a more comprehensive understanding of the implications of the canopy cover on
480 germination and early growth of tree seedlings, it is important to keep in mind that growing conditions
481 are influenced by complex interactions between air temperature, tissue temperature, and both
482 atmospheric and soil moisture (McLaughlin et al. 2017; Davis et al. 2019)(Table S3). Future research
483 should also monitor microclimate conditions before and after harvesting to better assess the effects of
484 forest management independent of site conditions. Additionally, we sampled the harshest topographical
485 position by focusing on south-facing aspects at a mid-slope position where incoming solar radiation is
486 highest. Further research on possible interaction effects of canopy cover and protected topographical
487 positions on microclimate conditions should also be conducted (Meineri et al. 2015; Finocchiaro et al.
488 2024).

489 Our results raise concern about the temperature conditions these regenerating plants will
490 experience. However, if the establishment of the first understory cohort is successful, the additional
491 shading by understory vegetation and advance regeneration will increase the amount of near surface
492 temperature reduction (Kovács et al. 2017; Prévosto et al. 2020). This raises concerns about a positive
493 feedback loop where the increase in temperature stress immediately after partial harvest slows or
494 prevents the establishment and growth of tree and other understory vegetation, leading to arrested
495 succession driven by continuing high temperatures and evaporative demand that further affect growth
496 and establishment of understory layers (Dey et al. 2019; Soto and Puettmann 2020).

497 7. IMPLICATIONS FOR MANAGEMENT

498 The tradeoffs discussed above can be used to inform management actions to mitigate climate
499 change effects and promote natural regeneration. Current climate change adaptation and forest

500 restoration treatments focus on reducing stand density for drought and fire resilience (Sohn et al. 2016;
501 Bottero et al. 2017; Stephens et al. 2020), creation of structural complexity (Puettmann et al. 2016;
502 Stephens et al. 2020), and recently promoting natural regeneration (Dey et al. 2019). Sampling across a
503 gradient of canopy cover allows our findings to be relevant to a wide variety of these treatments from
504 homogenous thinning prescriptions in even-aged stands to variable density treatments (Puettmann et
505 al. 2016; Franklin and Donato 2020). Thus, how managers may use the results of this study to mitigate
506 impacts on forest ecosystems will vary based on the management goal and the available resources. For
507 example, if the goal is to regenerate a cohort of shade-intolerant species and there are sufficient
508 resources to allow for multiple entries or treatments, managers may retain more overstory cover to
509 provide shading and temperature buffering during the early stages of seedling establishment. Once the
510 new cohort has established and has greater heat tolerance (Harper 1977), a larger portion of the
511 overstory could be removed to improve light availability (Ashton and Kelty 2018; Palik et al. 2020).

512 When there are fewer resources available for multiple entries, which is often the case for
513 federally managed lands in the western US, the results from this study can be used to guide post-harvest
514 canopy cover percentages based on topography and the desired understory composition. For example,
515 consider a project area with variable topography and a goal of creating structurally complex multi-aged
516 stands over half the area and on the other half regenerating stands focused on timber production
517 relying on regeneration from seed to achieve both goals. Results from this study suggest identifying the
518 harshest sites based on topographical condition for leaving higher post-harvest canopy cover. Due to the
519 lower light availability but buffered temperature conditions in the understory this would provide
520 opportunity for recruitment of shade-tolerant species, which are often less heat tolerant, eventually
521 resulting a structurally complex and diverse overstory (Kuehne and Puettmann 2008; Puettmann et al.
522 2016). When the desired future stand condition requires less overstory canopy cover, such as
523 establishment of shade-intolerant species for timber production, our results suggest avoiding south-

524 facing aspects for these prescriptions. Our results show that, on harsh topographic positions,
525 regenerating seedlings with little shading from overstory canopy cover are likely to experience heat
526 stress and are vulnerable to extreme heat events. Thus, results of this study can be used to mitigate
527 climate change impacts while achieving a variety of management goals by balancing the trade-offs
528 associated with overstory canopy cover with topographic conditions and the desired understory
529 condition and composition.

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539 9. COMPETING INTERESTS: The authors declare there are no competing interests.

540 10. DATA AVAILABILITY

541 Data generated or analyzed during this study are available from the corresponding author upon
542 reasonable request.

543 11. REFERENCES

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900

901 FIGURE CAPTIONS

902 **Figure 1:** Overall location of the study area within Oregon and locations of the 20 sensors within 11
903 harvest units that were part of the larger Big Blue Project area, and PRIMET and CENMET meteorological
904 stations (ESRI 2021, USDA Forest Service 2009). Figure was created using ArcGIS Pro 3.1.3 and
905 assembled from the following data sources: [USDA Forest Service](#), [HJ Andrews Experimental Forest and](#)
906 [LTER site](#). Base map from ESRI courtesy of Linn County, Bureau of Land Management, State of Oregon,
907 State of Oregon DOT, State of Oregon GEO, Esri Canada, Esri, HERE, Garmin, USGS, NGA, EPA, USDA,
908 and NPS

909 **Figure 2:** Estimated relationship and associated 95% confidence intervals between the three response
910 variables and canopy cover. To isolate the effect of canopy cover, all other fixed effects (see Table 1)
911 were held at their observed median values.

912 **Figure 3:** Estimated relationships between the three response variables and canopy cover, as measured
913 by densiometer, for 4 different climate scenarios and the presence/absence of smoke. Fitted lines were
914 plotted using a prediction dataset where all elevations and heat loads were held at their observed
915 median. For each climate scenario open-air temperature was held at the associated temperature. To
916 account for the effect of smoke incoming shortwave radiation was held at the value for the week of July
917 27th without the presence of smoke and the smoke used the value from the week of August 3rd (See
918 Table A2 for weekly values of each variable).

919

Table 1. Model parameters of the 3 different model types and 12 models fit in the analysis to address the two objectives. Each model type had 4 individual models with the same response variable. These four models only differed in which measurement of canopy cover was used (360° measurement from spherical densiometer and three LiDAR-derived canopy cover estimates for different time periods). AM time period corresponds to shade from 9am-12pm, PM time period corresponds to shade from 12pm-3pm, and AM + PM is 9am to. Refer to Figure A1 for how these variables were calculated. For each response variable, weather station screen height air temperature used in the models was summarized to match the response variable.

Objective	Response Variable	Model Type	Canopy Cover	Continuous Fixed effects	Random Effects
1	Weekly average daily maximum	Linear Mixed Model with AR1 Correlation Structure	360° AM PM AM + PM	1. Canopy Cover (%) 2. Open air temperature (°C)	
2	Presence/absence of Stress Degree Hours	Binomial Generalized Linear Mixed Model with Logit Link	360° AM PM AM + PM	3. Elevation (m) 4. Heat load 5. Weekly average daily accumulation of incoming shortwave radiation (J/m ² /week)	Sensor nested in harvest unit
2	Log of Weekly average daily accumulation of Stress Degree Hours	Linear Mixed Model with Log Transformation and AR1 Correlation Structure	360° AM PM AM + PM		

11 **Table 2.** Results of model comparison using Δ AICc and pseudo R^2 to evaluate which model and
 12 associated canopy cover variable better predicted each of the three response variables. The model with
 13 the lowest Δ AIC was then used (in bold). For each of the three response variables this was the 360°
 14 measurement from a convex spherical densiometer.

Objective	Response Variable	Model Type	Canopy Cover	AICc	Δ AIC	Pseudo R^2			
1	Weekly average daily maximum	Linear Mixed Model with AR1 Correlation Structure	360°	1084.67	0.00	0.82			
			AM + PM	1086.15	1.48	0.76			
			AM	1088.38	3.71	0.74			
1			PM	1091.72	7.05	0.76			
			2	Presence/absence of Stress Degree Hours	Binomial Generalized Linear Mixed Model with Logit Link	360°	146.59	0.00	0.72
						AM + PM	153.33	6.74	0.62
AM	153.72	7.13				0.62			
2	Log of Weekly average daily accumulation of Stress Degree Hours	Linear Mixed Model with Log Transformation and AR1 Correlation Structure	360°	339.85	0.00	0.50			
			AM + PM	353.26	13.44	0.26			
			AM	353.52	13.67	0.25			
2			PM	353.95	14.09	0.17			

15

17 **Table 3.** Estimates and 95% confidence intervals for relationships between the three response variables
 18 (weekly average daily maximum, presence/absence of SDH, and amount of SDH accumulated)
 19 associated with the two objectives. Objective column connects estimates to hypotheses and provides
 20 context to which relationship the estimate applied to. For the binomial GLMM estimates and confidence
 21 intervals
 22 were exponentiated from the link scale (log odds) to the odds scale

Objective	Model Type	Canopy Cover	Lower 95% CI	Estimate	Upper 95% CI
1	Change in mean weekly average daily maximum temperature at 2cm for 10 % change in canopy cover Linear Mixed Model with AR1 Correlation Structure	360°	0.41	1.32	2.24
		AM + PM	0.48	1.25	2.03
		AM	0.26	0.95	1.63
		PM	0.08	0.64	1.20
2	Factor for the multiplicative change in odds of accumulation of SDH for a 10% change in canopy cover Binomial Generalized Linear Mixed Model with Logit Link	360°	0.07	0.26	0.62
		AM + PM	0.12	0.48	1.48
		AM	0.13	0.53	1.51
		PM	0.27	0.65	1.59
2	Factor for the multiplicative change in median weekly average daily accumulation of SDH for a 10% change in canopy cover Linear Mixed Model with Log Transformation and AR1 Correlation Structure	360°	0.46	0.61	0.81
		AM + PM	0.68	0.96	1.36
		AM	0.70	0.91	1.18
		PM	0.84	1.06	1.33

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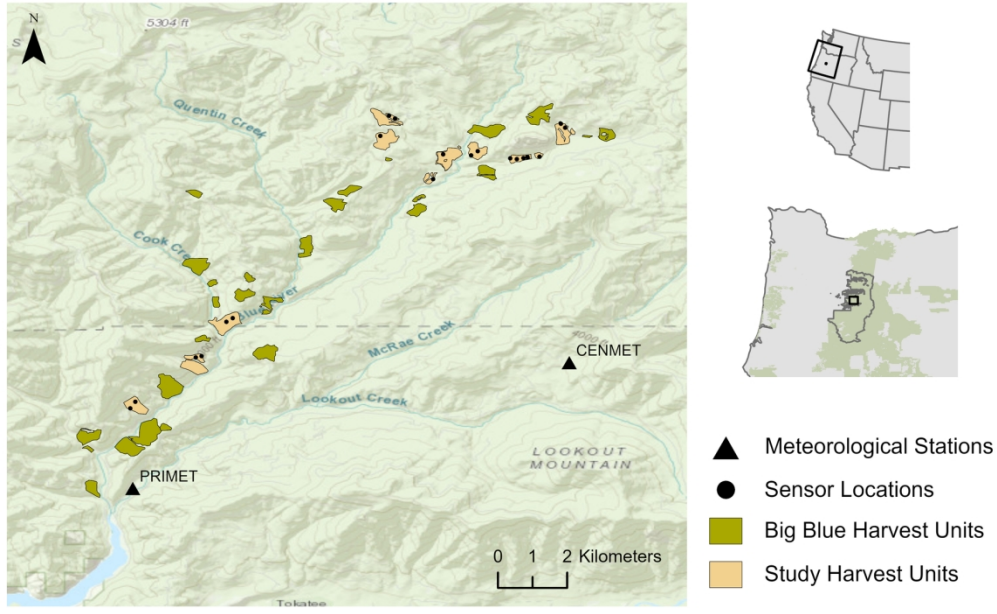


Figure 1: Overall location of the study area within Oregon and locations of the 20 sensors within 11 harvest units that were part of the larger Big Blue Project area, and PRIMET and CENMET meteorological stations (ESRI 2021, USDA Forest Service 2009). Figure was created using ArcGIS Pro 3.1.3 and assembled from the following data sources: USDA Forest Service, HJ Andrews Experimental Forest and LTER site. Base map from ESRI courtesy of Linn County, Bureau of Land Management, State of Oregon, State of Oregon DOT, State of Oregon GEO, Esri Canada, Esri, HERE, Garmin, USGS, NGA, EPA, USDA, and NPS

160x99mm (300 x 300 DPI)

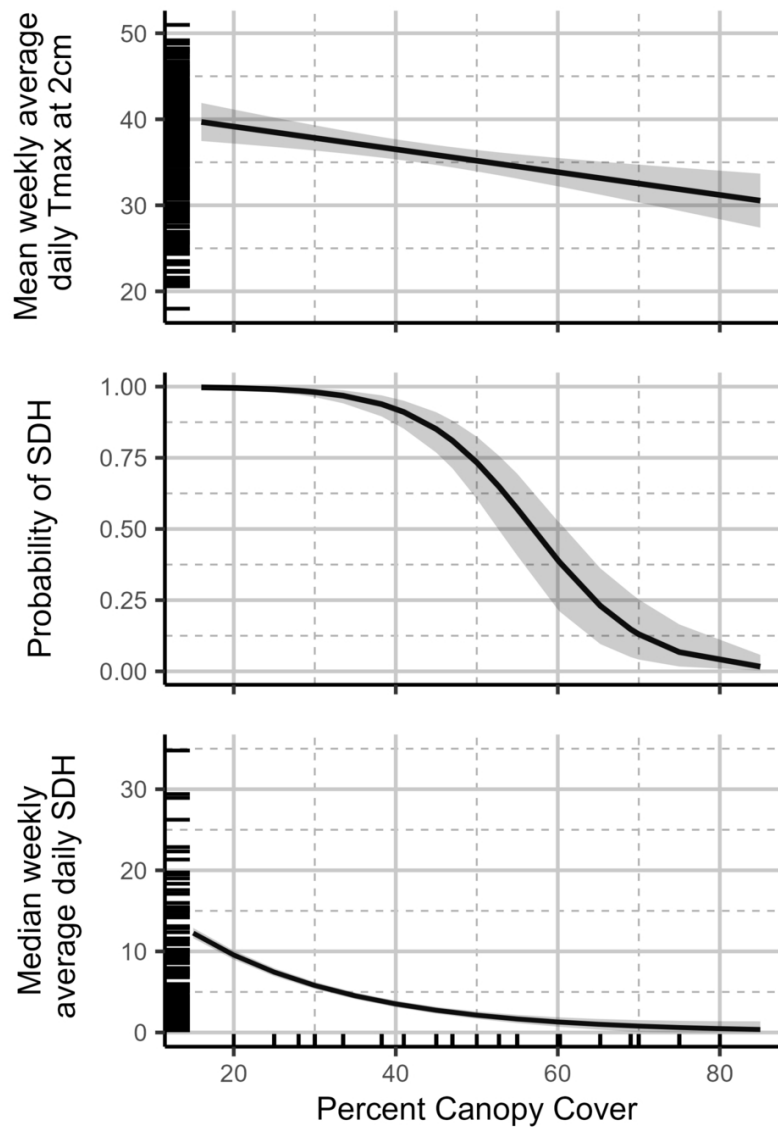


Figure 2: Estimated relationship and associated 95% confidence intervals between the three response variables and canopy cover. To isolate the effect of canopy cover, all other fixed effects (see Table 1) were held at their observed median values.

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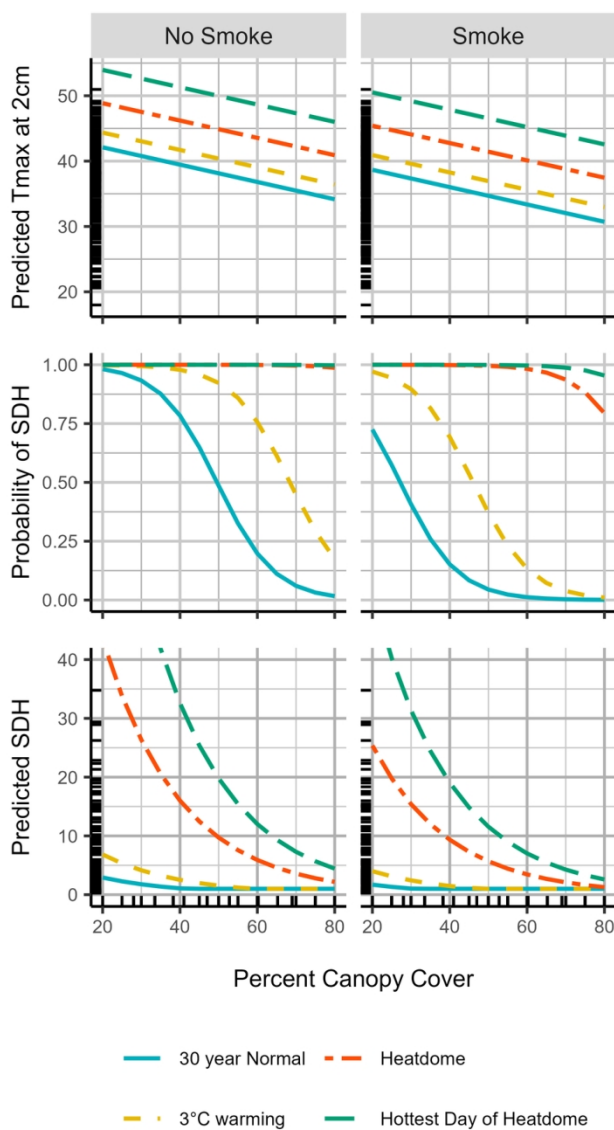


Figure 3: Estimated relationships between the three response variables and canopy cover, as measured by densiometer, for 4 different climate scenarios and the presence/absence of smoke. Fitted lines were plotted using a prediction dataset where all elevations and heat loads were held at their observed median. For each climate scenario open-air temperature was held at the associated temperature. To account for the effect of smoke incoming shortwave radiation was held at the value for the week of July 27th without the presence of smoke and the smoke used the value from the week of August 3rd (See Table A2 for weekly values of each variable).

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