

1 Title: Response of western Oregon stream temperatures to contemporary forest management

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35 ¹**Abstract:** A replicated before-after-control-impact study was used to test effectiveness of
36 Oregon’s riparian protection measures at minimizing increases in summer stream temperature
37 associated with timber harvest. Sites were located on private and state forest land. Practices on
38 private forests require riparian management areas around fish-bearing streams; state forest’s
39 prescriptions are similar but wider. Overall we found no change in maximum temperatures for
40 state forest streams while private sites increased pre-harvest to post-harvest on average by 0.7 °C
41 with an observed range of response from -0.9 to 2.5 °C. The observed increases are less than
42 changes observed with historic management practices. The observed changes in stream
43 temperature were most strongly correlated with shade levels measured before and after harvest.
44 Treatment reach length, stream gradient, and changes in the upstream reach stream temperature
45 were additionally useful in explaining treatment reach temperature change. Our models indicated
46 that maximum, mean, minimum, and diel fluctuations in summer stream temperature increased
47 with a reduction in shade, longer treatment reaches, and low gradient. Shade was best predicted
48 by riparian basal area and tree height. Findings suggest that riparian protection measures that
49 maintain higher shade such as the state forests were more likely to maintain stream temperatures
50 similar to control conditions.

51

52 Keywords: Riparian Buffer, Stream Temperature, Mixed-effects, Shade

53

54 **1. Introduction**

55 The Oregon Coast Range supports several cold-water fisheries (e.g. salmon, steelhead, cutthroat)
56 which are important to the region’s economy, culture, and recreational activities. These fish are
57 thermally adapted to specific water temperatures for various life stages such as egg and smolt
58 survival, spawning, and adult migration (Richter and Kolmes, 2005). Because forest
59 management can influence stream temperature regimes it is important to evaluate the
60 effectiveness of contemporary riparian management strategies to protect stream temperature.

¹ Abbreviations: AZ – Azimuth; BAPH – Basal Area per Hectare; ChannelBD – Channel Blowdown; CR – Crown Ratio; CT – Control reach temperature change standardized by reach length; EL – Elevation; FPA – Forest Practices Act; GR - Gradient; HS – Harvest Status; ODEQ – Oregon Department of Environmental Quality; OW - Owner; PlotBD – Plot Blowdown; PP – Private site post-harvest; RMA – Riparian Management Area; SH - Shade; TL – Treatment Reach Length; TPH – Trees per Hectare; WS – Watershed area

61
62 Stream temperature patterns are the result of several energy transfer processes and reflect both
63 the seasonal change in net solar radiation and the spatial and temporal changes in other energy
64 transfer processes including evaporation, convection, conduction, and advection (Brown 1969,
65 Caissie, 2006; Johnson, 2004). The magnitude and direction of these energy transfer processes
66 are influenced by atmospheric conditions (solar radiation, air temperature, wind speed, cloud
67 cover and relative humidity), basin level physical factors such as surrounding topography,
68 surface and groundwater flow, and streamside vegetation (Poole and Berman, 2001; Sinokrot and
69 Stefan, 1993). The amount of shade provided by streamside vegetation is perhaps the most
70 important single variable affecting summertime stream temperatures in forested environments
71 (Brown, 1969; Johnson and Jones, 2000; Danehy et al., 2005).

72
73 Historic forest management along streams often resulted in dramatic reductions in shade and
74 associated increases in stream temperature (Brown and Krygier, 1970; Levno and Rothacher,
75 1967; Harris, 1977; Holtby and Newcombe, 1982; Feller, 1981; Johnson and Jones, 2000; Moore
76 et al., 2005). Moore et al. (2005) summarize findings from several historic and contemporary
77 studies in the rain-dominated region of the Pacific Northwest. Reported changes in maximum
78 stream temperature for sites harvested without a riparian buffer ranged from 1.4-11.6°C. The
79 changes were considered to be a result of increased solar radiation at a time when harvesting did
80 not require riparian buffers, slopes were broadcast burned to the edge of the stream, and
81 equipment operation occurred in stream channels.

82
83 Varying levels of effectiveness for contemporary riparian buffers have been reported in the
84 literature. Moore et al. (2005) reported a range of 2.5-5.0°C for streams with riparian buffers in
85 the Pacific Northwest. The variability in stream temperature responses may be due in part to
86 different management practices. Within the reported literature sites differ according to harvest
87 strategy (clearcut harvest or different levels of thinning) and levels of riparian vegetation
88 retention (Moore et al. 2005). The variability in temperature responses may additionally be a
89 function of multiple biological and physical site factors that affect the energy transfer processes.
90 For example, while removal of vegetation can decrease shade it can also result in an increase in
91 summer low flows. This may result in an increase in cool groundwater that would moderate the

92 effects of a shade reduction on stream temperature (Moore and Wondzell, 2005). Stream
93 velocity and depth also influence the sensitivity of the stream to change (Brown, 1969). Danehy
94 et al. (2005), Isaak et al. (2010), and Isaak and Hubert (2001) found elevation to serve as a
95 predictor of stream temperature. Gomi et al. (2006) posited that valley azimuth may have
96 influenced the effectiveness of leave-tree buffers at intercepting incoming solar radiation.
97 Geomorphic features such as channel sediment deposits (Johnson, 2004) or channel reaches
98 scoured by debris flows (Levno and Rothacher, 1967) can also influence the sensitivity of the
99 stream to changes in temperature changes.

100

101 Determining the effects of contemporary harvest practices on stream temperature involves
102 detecting relatively small changes in stream temperature within a wide range of background
103 variability. Background variability is a function of several factors including basin size [Caissie,
104 2006; Lewis et al., 1999], microclimatic and geologic processes (Brosofske et al., 1997; Hawkins
105 et al., 1997; Kasahara and Wondzell, 2003), and annual and spatial hydrological variability
106 (Poole and Berman, 2001; Quinn and Wright-Stow, 2008). As such, longitudinal patterns can be
107 highly variable in small streams (Dent et al., 2008; Torgerson et al., 1999). Study and analysis
108 designs must therefore be able to separate this inherent variability from potential harvest effects.

109

110 Groom et al. (2011) evaluated changes in stream temperature following harvest in the Oregon
111 Coast Range. Their study design was developed to account for aforementioned spatial and
112 temporal variability. They reported a 40% increase in the probability that stream temperatures
113 would exceed 0.3 °C following timber harvest conducted according to Oregon harvest
114 regulations for small and medium fish-bearing streams. The frequency of temperature increases
115 did not surpass background levels for sites harvested according to state forest standards. While
116 their analysis allowed them to address a state water quality standard it hindered their ability to
117 provide estimates of the actual magnitude of change or to evaluate potential drivers of stream
118 temperature change (e.g., treatment reach length, shade, riparian and stream characteristics).

119

120 Quantification of the magnitude of stream temperature change allows for the comparison of
121 results to other studies and historical effects of harvest, and provides insight into background
122 variability. Also, the development of changes in riparian policy requires an understanding of how

123 management-controlled factors affect shade and temperature changes. Therefore the objectives
124 of this analysis are to:

- 125 1. Identify site physical and vegetative factors, including shade, that relate to stream
126 temperature change.
- 127 2. Determine the magnitude of stream temperature change that results from timber harvest.
- 128 3. Quantify riparian characteristics that predict shade retention after harvesting.

129 We expand our assessment of stream temperature beyond the water quality standard-focused
130 weekly maximum temperatures reported by Groom et al. (2011) and examine daily maximum,
131 minimum, and mean temperatures as well as diurnal fluctuation, in order to better capture the
132 spectrum of temperature changes following harvest. The implications of findings for this study
133 likely extend to other regions with similar physical and biological characteristics, stream
134 temperature concerns, cold-water fisheries, and prescriptive riparian zone protections such as
135 Idaho, Alaska, British Columbia, Washington, and California.

136

137 **2. Methods**

138 *2.1. Study location and design*

139 This study was conducted at 33 sites in the Oregon Coast Range (see Figure 1 in Dent et al.,
140 2008). Sites were situated along first- to third-order streams on 18 private and 15 state forest
141 sites dominated by Douglas fir (*Pseudotsuga menzeisii*) and red alder (*Alnus rubra*). Forest
142 stands were 50-70 years old and were fire- or harvest-regenerated. Openings were dominated by
143 shrubs such as vine maple (*Acer circinatum*), stink currant (*Ribes bracteosum*), salmonberry
144 (*Rubris spectabilis*), and devil's club (*Oplopanax horridus*).

145

146 The study design rendered a probabilistic sampling approach impractical due to the specificity of
147 site inclusion criteria and the resulting scarcity of sites. The criteria included an ability to collect
148 at least two years of pre-treatment and five years of post-treatment data at every site, minimum
149 treatment reach lengths of 300 m, and assurance that the upstream “control” reaches would
150 remain unharvested for the duration of the study. Streams needed to qualify under the Oregon
151 Forest Practices Act (FPA) as “Small” or “Medium” (mean annual streamflow < 57 or between
152 57 and 283 L/s, respectively), and streams needed to be free of recent impacts from debris
153 torrents and active beaver ponds. We obtained sites by requesting that industrial private and

154 state forest managers in the Oregon Coast Range provide ODF with a list of stream reaches that
155 met the criteria and would be harvested no sooner than 2004. An initial list of 130 stream
156 reaches was reduced to 36 (three more were subsequently dropped due to changes in harvest
157 plans) that met study design constraints. Assuming selected sites were geographically
158 representative (all available sites that met selection criteria were included in the study),
159 inferential scope of study results pertain to first- to third-order streams within the Coast Range,
160 on 50- to 70-year-old non-federal forestlands primarily managed for timber production that lack
161 recent debris torrent or beaver disturbance. While there was an initial attempt to exclude sites
162 with beaver activity, a beaver dam ponded 220 m of the 1.16 km treatment reach for site 7801
163 during the first and second post-harvest years.

164
165 A site's control reach was located immediately upstream of its treatment reach (Figure 1). The
166 control reaches were continuously forested to a perpendicular slope distance of at least 60 m
167 from the average annual high water level. Reach lengths varied from 137 m to 1,829 m with
168 means of 276 m and 684 m for the control and treatment reaches respectively.

169

170 *2.2. Treatments*

171 *Forest Practices Act On Private Sites:* Eighteen sites were established on private forest streams.
172 Sites were harvested following contemporary FPA standards which require riparian buffers along
173 fish-bearing streams to protect stream temperature, provide future large wood for streams, and
174 retain other ecological services (Oregon Department of Forestry, 2007). Under the FPA, Coast
175 Range RMAs are 15 and 21 m wide around small and medium fish-bearing streams, respectively.
176 Both small and medium streams may not be harvested within a 6 m zone immediately adjacent to
177 the stream. Harvesting is allowed in the remaining RMA to a minimum basal area of 10.0 (small
178 streams) and 22.9 (medium streams) m²/ha.

179

180 *Oregon State Forest Management Plan (FMP) on State Sites:* Land administered directly by
181 Oregon Department of Forestry (state forests) is managed under the FMP for multiple resource
182 objectives including timber production (e.g., recreation, wildlife) and require riparian protections
183 that exceed FPA minimum values. State RMAs are 52 m wide for all fish-bearing streams, with
184 an 8-m no cut zone. Limited harvest is allowed within 30 m of the stream only to create mature

185 forest conditions. Harvest operations within this zone must maintain 124 trees per hectare and a
186 25% Stand Density Index (Oregon Department of Forestry, 2001). If mature forest conditions
187 exist or will develop in a timely manner without management, then no harvest is allowed within
188 30 m of the stream. Additional tree retentions of 25-111 conifer trees and snags/hectare are
189 required between 30 and 52 m.

190

191 *2.3. Data Collection*

192 Optic Stowaway Temp and HOBO Water Temp Pro data loggers (Onset Computer Corporation,
193 Bourne, Massachusetts) were annually deployed at three stations within each site beginning in
194 2002 or 2003 (Figure 1). Station 1 was located at the upstream end of the control reach, Station
195 2 was located at the downstream end of the control reach and the upstream end of the treatment
196 reach (i.e. shared logger) and Station 3 was situated at the downstream end of the treatment
197 reach. Temperature loggers were deployed in shaded locations where stream flow was relatively
198 constant at a reliable summer depth and within a well-mixed water column. Logger accuracy
199 was checked prior to installation, in the field, and following retrieval with National Institute for
200 Standards and Technology-calibrated digital thermometers according to Oregon Watershed
201 Enhancement Board (1999) stream temperature protocol. Although both types of loggers are
202 listed as ± 0.2 °C accuracy, we found that for over 500 pre- and post- deployment assessments, in
203 only two instances did loggers register errors of > 0.1 °C. Daily temperatures that exhibited
204 increases in diel fluctuation and increases and decreases in daily maximum and minimum
205 temperatures that were not reflected in other probes during the same year or at the same location
206 during other years were excluded from the analysis.

207

208 Channel data were collected at 60 m intervals within each reach. Data included wetted width,
209 bankfull width, thalweg depth, and stream gradient according to the protocol described by
210 Kaufmann and Robison (1998). Stream shade was quantified at each 60 m interval using a self-
211 leveling fisheye lens digital camera (Valverde and Silvertown, 1997). Shade values were
212 measured once pre-harvest and once post-harvest. Fish-eye photographs were taken in the
213 middle of the stream, 1 m above the water level, and oriented due north. Photographs at this
214 height may slightly underestimate stream shading due to shading from the banks, lower-growing
215 vegetation, and channel wood (Davis-Colley and Quinn, 1998). An effort was made to take

216 photos when the sun would not be in the picture (e.g. overcast conditions, when the sun was
217 below the topographic horizon) and under dry conditions. Shade values were calculated from the
218 photographs using HemiView™ 2.1 software (Delta-T Devices, Cambridge, UK) as one minus
219 the June 30 Global Site Factor (1 -GSF). The GSF is the proportion of both direct and diffuse
220 energy under a plant canopy relative to the available direct and diffuse energy for the given site's
221 latitude/longitude. Shade and gradient values were averaged for each reach.

222
223 Vegetation data were collected in four 152 by 52 m plots on either side of a study stream in the
224 treatment and control reach (Figure 1). Plots were centered midway along each reach and altered
225 in layout to accommodate stream nonlinearity. Vegetation plot data describe understory,
226 overstory, downed wood, blowdown, and snag characteristics. The original purpose of including
227 extensive vegetation plot data collection was to assess large wood recruitment, shade, and
228 riparian structure following timber harvest. For this analysis we use a portion of the available
229 riparian structure data (e.g. blow down, tree heights, basal area) that both influence stream
230 temperature and relate to timber management. Within each plot all living trees with a diameter
231 at breast height (DBH, or diameter at 1.4 m above ground level) > 14 cm were tallied by species
232 and the distance from those trees to the bankfull edge of the stream recorded. Height and live
233 crown ratio were additionally measured for 20% of the trees. Data were collected in all four
234 vegetation plots per site pre-harvest and re-measured in treatment plot or plots (if one or both
235 stream sides were harvested, respectively) post-harvest. Blowdown was quantified in all plots
236 post-harvest.

237
238 We determined stream elevation (elevation at the Station 3 logger), treatment reach azimuth, and
239 watershed area from a geographic information system (GIS). We determined study site elevation
240 from examining GPS-determined logger locations against 10-m digital elevation models (USGS
241 National Mapping Program; overall absolute vertical root mean square error of 2.44 m).

242 Azimuth was calculated using GPS locations of treatment reach upstream and downstream
243 loggers. Since an east-west valley azimuth is expected to deliver the most solar radiation to a
244 stream (Gomi et al., 2006; Sridhar et al., 2004), we “folded” (rendered equivalent) the azimuth
245 north to south by subtracting 180 degrees (Hawkins et al., 1997; Bartholow, 1989), then folded
246 the azimuth east to west by subtracting an additional 90 degrees . The result represents a 0 – 90°

247 deviation from either east or west. Watershed area was calculated using the Spatial Analyst
248 extension's Watershed tool for ArcGIS 9.3.1 (Earth Systems Research Institute, Redlands, CA).
249 The tool relies on the same 10-m DEM as was used for elevation. Watershed area was calculated
250 at the downstream end of the treatment reach (Station 3).

251

252 *2.4. Analysis*

253 We limited this analysis to include all pre-harvest data and data from the first and second post-
254 harvest years. This decision was made due to a staggered harvest schedule and some prolonged
255 pre-harvest periods resulting in incomplete five-year post-harvest data. Data collection included
256 hourly stream temperature data (collected annually between July 1 and September 15) and both
257 channel data (for a complete list see Dent et al., 2008) and riparian vegetation data (overstory
258 and understory) during the first years pre-harvest and post-harvest.

259

260 We examined our stream temperature, riparian vegetation, and channel data with two distinct
261 analyses. We constructed one analysis to functionally understand processes influencing stream
262 temperature, which addressed the first two of our three objectives. This analysis involved four
263 measures of temperature as separate dependent variables and stream, shade, and riparian
264 characteristics as independent variables. We developed several models for this analysis to
265 determine a reasonable fit for the data as well as to obtain magnitude of effects estimates. To
266 address the third objective, we conducted another analysis that examined several riparian
267 vegetation metrics to determine which best related to observed stream shade values post-harvest.
268 For both analyses we plotted and assessed potential explanatory variables by examining
269 Pearson's correlations and variance inflation factors (VIF) to determine degrees of
270 multicollinearity. We retained all temperature and shade analysis variables as VIF scores were
271 all < 10 , indicating a lack of multicollinearity (temperature maximum = 1.76, shade maximum =
272 2.50; Neter et al., 1996). Prior to analysis we centered all independent variables by subtracting
273 the variable mean from all values. We assessed dependent variables (change in treatment reach
274 temperature for the four temperature analyses and Shade for the shade analysis) graphically to
275 determine whether transformations appeared appropriate.

276

277 2.4.1 Analysis I: Stream Temperature

278 We summarized hourly stream temperature data to provide daily maximum, mean, minimum,
279 and fluctuation (maximum-minimum) values for each data logger. Our analysis objectives 1 and
280 2 concerned detecting changes in stream temperature due to site factors including harvest. We
281 therefore defined the response variable as the difference between treatment reach Stations 2 and
282 3. To reduce analysis complexity we computed the average of this difference over a forty day
283 period for each year (July 23 to August 15). This represents the time frame when we had the
284 most functional loggers recording temperatures during a central portion of the summer months
285 when maximum temperatures are observed in the Oregon Coast Range. Averaging of
286 temperature differences removed the variability associated with temperature differences from
287 shorter time periods, such as single days. Temperature changes observed in this study therefore
288 reflect prolonged alterations in stream conditions. The stream temperature analyses utilize the
289 averaged differences in daily maximum, mean, minimum, and fluctuation values as separate
290 dependent variables and we refer to these metrics as Maximum, Mean, Minimum, and Diel
291 Fluctuation.

292
293 Independent variables used for modeling temperature included treatment reach length (TL),
294 average treatment reach shade (SH), elevation (EL), average treatment reach gradient (GR),
295 treatment reach valley degree deviation from east or west (AZ), change in control reach
296 temperature standardized by control reach length (CT; ControlMax, ControlMean, ControlMin,
297 ControlFluc), and watershed area calculated at Station 3 (WS). We also included state and
298 private ownership (OW), the harvest status or whether a temperature measurement occurred
299 during a pre-harvest or post-harvest year (HS), or a variable that indicated temperature data were
300 from private forest during post-harvest (PP; included due to its importance in Groom et al.,
301 2011). Except for shade and harvest status, all other explanatory variable values were static over
302 all years. The variables TL and CT were included in all temperature models considered except
303 an intercept model (Intercept) and one that only included CT (Upstream, Table 1).

304
305 Our modeling approach involved first selecting an appropriate random-effects model structure,
306 followed by an evaluation of potential explanatory models of interest. We modeled the
307 temperature data using mixed-effects linear regression (Pinheiro and Bates 2000). We selected
308 this technique due to a concern of potential data non-independence within sites and within

309 specific years. To determine the appropriate random effects structure for each of the
310 temperature analyses we created an over-fit “beyond optimal” linear model (BO; Zuur et al.,
311 2009) that additively included all independent variables and accounted for as much of the fixed-
312 effects variation as possible. We then compared different random-effects structures that used
313 this fixed-effects model structure. The random effect structures differed according to data
314 groupings and random-effects parameterization. The models were fit using Maximum
315 Likelihood estimation to enable comparison of model AIC values (Zuur et al., 2009). Random
316 effects structures included a generalized least squares model (no grouping), a random intercept
317 model grouped by site, another by year (2002 through 2008), and a third by a cross of site and
318 year. We examined grouping data by year to potentially account for inevitable interannual
319 differences in air temperature, stream flow, and precipitation (Table 2). Three additional models
320 included a random intercept and slope for control temperature change and grouped data
321 respectively by site, year, and a cross of site and year. The best-supported (smallest AIC)
322 Beyond-Optimal model for all four temperature metrics grouped data by site and fit random
323 effects parameters for intercept and control reach temperature change (respectively the control
324 change in Maximum, Minimum, Mean, or Diel Fluctuation).

325
326 Once we selected an appropriate random-effects parameterization we altered the fixed effects
327 and fit a suite of 18 explanatory models (Table 1). The models were constructed a priori to
328 account for variation in stream temperature according to the stream’s in-channel, riparian, and/or
329 shade characteristics (Objective 1, Table 1 Maximum Rank models 1 through 5, 12 through 14,
330 16, and 17). To obtain temperature change magnitude estimates as a function of timber harvest
331 we included models that specified harvest status (pre- or post-harvest), ownership, or a
332 combination (Objective 2, Maximum Rank models 7 through 10). We included two more
333 models, BO, and Intercept, to serve as overfitting and underfitting extremes. We included
334 models Upstream and Upstream_TRlength to verify that CT and TL were indeed assisting model
335 fits. Models for comparison were fit using Maximum Likelihood estimation to enable
336 comparison of model AIC values; best-performing models were re-fit using Restricted Maximum
337 Likelihood to reduce bias in parameter estimates (Zuur et al., 2009). We compared model
338 weights (ω) to determine the probability that a model or subset of models represented or included
339 the best model of the set (Burnham and Anderson, 2002).

340

341 2.4.2. Analysis II: Shade

342 Shade analysis variables from the vegetation plots included trees per hectare (TPH), tree height
343 (Height), live crown ratio (CR), vegetation plot blowdown (PlotBD), and basal area per hectare
344 (BAPH). These variables were calculated by using vegetation plot data from the edge of the
345 bank to a perpendicular distance of 30 m, a distance at which tree canopies have likely ceased to
346 influence stream shade during daily periods of the greatest radiation intensity (mean measured
347 tree height = 25.7 m). We obtained site values for these variables by calculating mean values for
348 the two treatment reach plots. In instances where only a single side of the treatment reach was
349 harvested we calculated the mean TPH, BAPH, and CR values from the harvested side's post-
350 harvest values and the unharvested side's pre-harvest values. Height and CR values were
351 obtained from pre-harvest data. Non-vegetation plot variables included the number of riparian
352 banks harvested (Nsides), buffer width (BuffWidth), and channel blowdown (ChannelBD). We
353 quantified BuffWidth as the perpendicular distance from the stream bank to the first stump
354 encountered within 10 m of the observer, measured every 60 m along the treatment reach.
355 ChannelBD represented a tally of treatment reach within- or above-channel blowdown pieces >
356 15 cm diameter.

357

358 Shade data were logit transformed due to observed skewness. We performed a linear regression
359 analysis of shade data (n=33) and compared small-sample AIC values (AIC_c, Hurvich and Tsai,
360 1998) to determine relative model performance among 8 *a priori* models (Table 3). Models were
361 comprised of variable combinations we believed could potentially describe variations in
362 observed post-harvest site shade levels. Allen and Dent (2001) found that riparian basal area and
363 live crown ratio served as useful shade predictor variables for east-west flowing streams in the
364 Oregon Coast range. DeWalle's (2010) riparian shade model found combinations of specific tree
365 height, density, and basal area maximized stream shade. We therefore created several models
366 that contained combinations of BAPH, CR, TPH, and Height. At a more coarse level of analysis,
367 we anticipated that buffer width and the number of stream sides harvested might serve as
368 reasonable predictors of shade (model BuffWidth). An additional model anticipated that shade
369 values would be predicted by the quantity of blowdown both across the channel and within
370 vegetation plots (model Blowdown).

371

372 *2.4.3. Analysis assessment*

373 For both the stream temperature and shade analyses we examined q-q normal and residual plots
374 to verify that the best-performing models met their respective required statistical assumptions.
375 Linear models assume a linear relationship between independent and dependent variables,
376 independence of errors, a constant variance and normality of errors. Linear mixed effects models
377 additionally assume 1) within-group errors are independent and identically, normally, distributed
378 with a mean of 0 and a variance of σ^2 , and are independent of random effects; and 2) random
379 effects are normally distributed, with a mean equal to zero and a homogeneous random effects
380 covariance matrix (Pinheiro and Bates, 2000). Visual assessment of residual and q-q plots
381 indicated that the assumptions of within-group normality were plausible and within-group
382 residuals were symmetrical around zero. Within-group variances did not appear equal (not
383 surprisingly, given that the maximum number of points per site was 6 with a mean of 4 [Pinheiro
384 and Bates, 2000]); at the same time no site appeared to behave as an obvious outlier. Plots of
385 observed vs. fitted values indicated that the linear models agreed with observed values. For daily
386 temperature Maximum, Mean, and Diel Fluctuation, three values from two sites appeared to be
387 outliers, while we found one potential outlier for Minimum temperatures.

388

389 For the second assumption, normal q-q plots for the intercept and Control random effects
390 provided no indication of non-normality. Normal q-q plots of Maximum, Mean, and Minimum
391 intercept values revealed two potential outliers out of 33; Diel Fluctuation exhibited three. For
392 the CT random effect Maximum, Minimum, and Diel Fluctuation plots we found one outlier
393 apiece and zero for Mean. Homogeneity of random effects by group was difficult to assess due
394 to the general paucity of data for each site.

395

396

397 **3. Results**

398 *3.1. Analysis I: Stream Temperature*

399 Our ranking of models explaining observed temperature changes indicated that shade was
400 critical, with all other variables marginally improving model fit (Table 1). For change in daily
401 temperature Maximum, Mean, and Fluctuation the best-supported six models are those that

402 included SH; models that lacked shade exhibited a substantial drop in explanatory power
403 (increase in ΔAIC by > 13). We found little support for non-shade models as the cumulative
404 model weight for models that included shade was effectively 1.0. For changes in the Maximum
405 and Mean treatment reach temperatures the model with shade and gradient (Grad_Shade) and
406 gradient, shade, and watershed area (ShadeGradWS) performed best. The lowest-AIC model for
407 Diel Fluctuation was the overparameterized model BO. However, we interpret a more
408 parsimonious model El_Shade as better supported as it performed virtually as well ($\Delta AIC = 0.3$;
409 Burnham and Anderson 2001) and note that five other models performed similarly ($\Delta AIC < 2$).

410
411 We do not believe that our models described Minimum temperature behavior well (Table 1).
412 The best-supported models for Minimum temperatures were StreamShade, Grad_Shade, and a
413 model with an indicator variable for private sites post-harvest (Harvest_Private, combined
414 weight = 0.59), with StreamShade receiving the most support. Shade appeared to be important;
415 those models that contained SH provided a combined model weight of 0.74. However, the
416 indicator variable that distinguished private sites post-harvest from all other observations (PP)
417 explained the variation nearly as well as models that included SH (model weight for
418 Harvest_Private = 0.15) while the remaining predictor variables improved model fit marginally.

419
420 Parameter estimates for the best-supported models retained directionality but contributed to
421 model fit differently across temperature metrics (Table 4). For Maximum and Mean
422 temperatures, the variable CT appeared to account for expected changes in the treatment reach
423 given an observed temperature increase or decrease in the control reach. The decrease in
424 temperature between Stations 1 and 2 on average resulted in an increase in temperature between
425 Stations 2 and 3 and vice versa (Figure 2). CT was not a significant parameter for Minimum
426 temperatures or Diel Fluctuation. The parameters GR and EL were not strongly supported
427 (Table 1) and exhibited overlap in their estimates at the extremes of observed values (Figure 2).
428 The directionality of parameter TR indicated that streams warmed with increasing treatment
429 reach length. We estimated an increase in Maximum and Minimum temperatures of 0.73 and
430 0.59 °C per km, respectively (Figure 2). Low SH values were associated with temperature
431 increases in all models (Table 4).

432

433 Among explanatory variables, shade exhibited the greatest potential to alter stream temperatures
434 with minimum shade levels producing a predicted increase of ~ 2 °C and maximum shade levels
435 a temperature decrease of ~ -1 °C (Figure 2). We generally observed an increase in Maximum
436 temperatures pre-harvest to post-harvest for sites that exhibited an absolute change in shade of $>$
437 6%; otherwise, directionality appears to fluctuate (Figure 3).

438
439 State and private sites exhibited similarities in the mean and range of their site values for
440 gradient, treatment reach length, watershed area, crown ratio, wetted width, and pre-harvest
441 vegetation metrics (Table 5). Following harvest, Maximum temperatures at private sites
442 increased relative to state sites on average by 0.71 °C (Table 6, coefficient of PP, 95% CI = 0.51,
443 0.92). Similarly, Mean temperatures increased by 0.37 °C (0.24, 0.50), Minimum temperatures
444 by 0.13 °C (0.03, 0.23), and Diel Fluctuation increased by 0.58 °C (0.41, 0.75). This increase
445 appears to coincide with a decline in shade for some private sites (Figure 4). Harvest on state
446 sites did not produce a temperature change signal that differed from pre-harvest background
447 levels. Models for Maximum, Mean, and Diel Fluctuation that considered state harvest to have
448 negligible impact on stream temperature (Harvest_Private) received more support than those that
449 considered it an influence (model Harvest_S_P; Table 1) with an observed Δ AIC increase
450 respectively of 25.8, 16.2, and 5.9. The model which considered the variables PP and TL and
451 their interaction (Harvest_PXLength) generally had greater explanatory power (lower Δ AIC)
452 than the Harvest_Private models. However, the Harvest_PXLength model exhibited unexpected
453 parameter estimates, indicating greater increases in private stream temperatures along shorter
454 treatment reaches. A plot of post-harvest shade values against treatment reach length (Figure
455 5A) suggests this relationship may be due to shade loss along the shorter private reaches. This in
456 turn may be related to the presence of taller trees in the shorter reaches (Figure 5B) or lower
457 post-harvest basal area along the shorter reaches (Figure 5C).

458
459 A comparison of within-site changes in Maximum temperatures pre-harvest to post-harvest
460 indicated an overall increase in private site temperatures while observed changes at state sites
461 were as frequently positive as negative (Figure 6). The average of Maximum state site changes =
462 0.0 °C (range = -0.89 °C to 2.27 °C). Observed Maximum temperature changes at private sites
463 averaged 0.73 °C (range = -0.87 °C to 2.50 °C), and exhibit a greater frequency of post-harvest

464 increases from 0.5 °C to 2.5 °C compared to state sites. We repeated this comparison while
465 controlling for the effects of control reach temperature change, treatment reach length, and
466 gradient by plotting differences in partial residuals from the Maximum temperature model
467 Grad_Shade (each datum = model residuals + predicted effect of Shade). We found that state
468 site differences became less extreme for positive increases (<1.5 °C) while private comparisons
469 appeared to occupy the same range of responses (Figure 6).

470

471 3.2. Analysis II: Shade

472 Correlations among shade variables were greatest for comparisons with BAPH. Sites with
473 higher stocking levels, wider uncut buffers, or fewer stream banks harvested had greater basal
474 area. The variable BAPH exhibited a positive Pearson's correlation with TPH (0.54) and Width
475 (0.47) and a negative correlation with NSides (-0.55). Variable combinations Height and CR
476 were negatively correlated (-0.30) as were TPH and NSides (-0.52).

477

478 The shade model BasalXHeight which included parameters for basal area per hectare (BAPH),
479 tree height, and their interaction was best-supported (Table 3; $\Delta AIC_c = 0.0$, $n = 33$, $R^2 = 0.69$,
480 $\beta_{intercept} = 1.795$, $\beta_{BAPH} = 3.100e^{-2}$, $\beta_{Height} = -6.250e^{-2}$, $\beta_{BAPA*Height} = -4.680e^{-4}$,
481 model $p < 0.001$) with an AIC_c distance of 10.7 between itself and the next-best supported
482 model, BasalHeight (additive parameters BAPH and tree height). Its model weight ($\omega = 1.00$)
483 indicated strong relative support for this model and virtually no support for the remaining
484 models. The model BasalXHeight predicted greater shade coverage with shorter trees and more
485 basal area per hectare. An interpretation of the interaction term suggests that the effect of tree
486 height is most acute for riparian areas with greater basal area. Residual, normal q-q, and
487 standardized residual plots of the model parameterized by BAPH (BasalArea) indicated that
488 certain points exhibited substantial leverage to the extent that a linear model may not be
489 appropriate. Examination of paired variable plots suggested that plot blowdown may have
490 influenced the leveraging outliers. A post hoc linear model that included plot blowdown in
491 additional to basal area per hectare, tree height, and their interaction resolved the model fit issues
492 and fit the data better than BasalXHeight ($AIC_c = 7.5$ lower than for BasalArea, $R^2 = 0.75$).
493 However, site 5354 appeared responsible for contributing the blowdown effect; with its removal
494 the interaction term in BasalXHeight ceased to confer statistical significance ($p = 0.18$) and

495 BasalHeight graphically fits the data adequately ($R^2 = 0.70$). We did not find a relationship
496 between hardwood predominance among sites and tree height; the Pearson's correlation between
497 average percent hardwood at a site within 30 m of the stream and Height was 0.083 ($t = 0.466$, df
498 $= 31$, $p = 0.644$). An alternate shade model, BufferWidth (variables: number of sides harvested,
499 buffer width, and their interaction; Table 3), contained no vegetation plot variable information,
500 only the buffer width and number of riparian sides harvested. Its ΔAIC_c value distanced it from
501 the best-supported model yet it accounted for almost 50% of data variation ($n = 33$, $R^2 = 0.477$,
502 $\beta_{\text{intercept}} = 2.294$, $\beta_{\text{NSides}} = -0.699$, $\beta_{\text{BuffWidth}} = -4.470e^{-3}$, $\beta_{\text{NSides*BuffWidth}} = 1.230e^{-2}$,
503 model $p < 0.001$).

504
505 Private site shade values appeared to decrease pre-harvest to post-harvest. Private post-harvest
506 shade values differed from pre-harvest values (mean change in Shade from 85% to 78%, $n_{\text{Private}} =$
507 18 , $df = 17$, paired $t = -3.678$, $p = 0.002$); however, no difference was found for state site shade
508 values pre-harvest to post-harvest (mean change in Shade from 90% to 89%, $n_{\text{State}} = 15$ $df = 14$,
509 paired $t = -1.150$, $p = 0.269$). We did not find evidence that shade differed if one or both banks
510 were harvested for private sites ($n_{\text{SingleSide}} = 4$, $n_{\text{TwoSides}} = 14$, $df = 7.589$, $t = 1.978$, $p = 0.085$)
511 although the sample size for single sided harvests was low. Similarly, private site shade values
512 did not appear to differ between Medium or Small streams ($n_{\text{Small}} = 4$, $n_{\text{Medium}} = 14$, $df = 3.595$, t
513 $= -1.345$, $p = 0.257$).

514 515 **4. Discussion**

516 We estimated the magnitude of temperature changes by examining differences between pre- and
517 post-harvest summer maximum temperatures, evaluating predicted variable contributions to
518 temperature change, and comparing estimates of temperature change associated with private and
519 state forests. Maximum, Average, Minimum, and Diel Fluctuation stream temperatures increased
520 as a consequence of timber harvest on private forests. State forest stream temperature patterns
521 remained similar before and after harvest. The increases on private sites coincide with a decline
522 in shade due to timber harvest. In turn, the best predictor of post-harvest shade was a model
523 including basal area within 30 m of a stream, tree height, and their interaction.

524 525 4.1. Shade and Temperature Change

526 A primary driver of changes in stream temperature was stream shade. For the four temperature
527 values we examined, the mixed-effects models that included shade outperformed models that
528 lacked shade. The best-supported shade models indicated that the lowest-observed shade value
529 of 50% is associated with a predicted increase in Maximum stream temperatures by as much as 2
530 °C and for Minimum temperatures as little as 0.3 °C. At the greatest observed shade levels
531 (96%) the predicted response for Maximum and Minimum temperatures was -0.7 °C and -0.1 °C
532 respectively. This range of estimated temperature responses to site shade values is similar to
533 results from a manipulative experiment by Johnson (2004). She determined that for a 200 m
534 bedrock reach the temperature difference between 100% (artificial) shading and full exposure
535 stream temperatures differed by about 4 °C. Similar to our regression estimates of negative
536 Maximum stream temperatures at high shade values (-0.66 °C at 96% shade), Johnson found a
537 decline of about 1 °C in maximum stream temperatures with 100% shade.

538
539 Other shade model parameter estimates indicated that stream temperatures were expected to
540 increase with greater treatment reach lengths and low gradients (Table 4). The finding for
541 treatment reach length is consistent with Cassie (2006), who reported general stream temperature
542 increases with distance downstream. Our negative estimate for the coefficient of gradient
543 indicated that temperature increases over the length of our treatment reaches were less for steeper
544 reaches, a result that is echoed in Danehy et al. (2005), and that Subehi et al. (2009) attribute to
545 the reduced residence time of water within the stream. An alternative explanation is that more
546 frequent hyporheic exchange in steeper streams with step-pool morphologies may moderate
547 stream temperatures (Anderson et al. 2005). Our negative relationship between temperature
548 change in the control reach vs. the treatment reach was graphically described by Dent et al.
549 (2008). They found that an abrupt temperature increase in the control reach was generally
550 accompanied by an opposite change in the pre-harvest treatment reach. We suspect that local
551 hydrological conditions at one of the control probe stations, resulting in locally warmer or cooler
552 water temperatures (Bilby, 1984), could produced this temperature pattern. The temperature of
553 the downstream station would not reflect this condition and would therefore appear to reverse the
554 increase or decrease in temperature observed in the control reach.

555

556 4.2. Magnitude of Temperature Changes

557 The average increase in Maximum temperatures on private sites following timber harvest was
558 estimated to be 0.7 °C. However, sites did not all behave similarly; some decreased in
559 temperature while others exhibited higher increases. Although Mean, Minimum, and Diel
560 Fluctuation temperatures also increased overall post-harvest, they similarly exhibited variability
561 in response.

562
563 Overall, we did not observe a temperature increase on state sites as a result of timber harvest.
564 While some changes in stream temperature were observed, state forest treatment effects were not
565 substantial enough to support modeling them as differing from pre-harvest conditions. We
566 interpret these results and the general lack of observed changes in stream temperature for state
567 forest to indicate that treatment buffer widths and conditions on state forest sites were generally
568 sufficient to protect against timber harvest-related increases in stream temperature. Even so, two
569 state sites registered temperature changes between a pre-harvest year and a post-harvest year of >
570 2.3 °C; after controlling for site factors the increases were < 2 °C.

571
572 Our stream temperature models were parameterized using channel and local riparian conditions.
573 We did not examine variables outside of study reaches such as the influence of proportion of
574 watershed harvested. Solar exposure of clear-cut soils may affect evapotranspiration and soil
575 temperatures (Kim and Ek, 1995). St-Hilaire et al. (2000) improved empirical stream
576 temperature model fits by including the effect of solar exposure on soil temperatures. Hewlett
577 and Fortson (1982) found that despite the presence of a 10 to 15 m wide riparian buffer along
578 their study site in Georgia, USA, their study stream temperatures increased by more than 11 °C,
579 much more than the authors estimated would occur under conditions of complete riparian zone
580 removal. The authors suspected, based on ancillary data, that effluent groundwater may have
581 been warmed in the exposed areas and then flowed into their study stream. However, a follow-up
582 study on the same site with a more substantial riparian buffer has found no temperature increase
583 (Dr. Rhett Jackson, Univ. of Georgia, personal communication). Bourque and Pomeroy (2001)
584 found a relationship between stream temperature gains and proportions of upstream catchments
585 harvested. Pollock et al. (2009) similarly found a relationship between proportion of a watershed
586 harvested and maximum stream temperatures, although Ice et al. (2010) argue that these results
587 are best explained by riparian conditions such as shade and channel conditions. Our models

588 appear to have explained a substantial portion of the observed variability in stream temperature
589 and relationships to timber harvest; however, these factors may have played an additional role in
590 influencing our sites' stream temperatures.

591
592 The magnitudes of change in observed stream temperatures reported for this study are similar to
593 findings from other studies which evaluated contemporary harvest practices with riparian buffers
594 and substantially lower than values associated with older harvest practices or harvesting without
595 buffers. Levno and Rothacher (1967) found maximum temperature increases of 2.23 °C after
596 harvest but before stream scouring (logging debris partially shaded the stream) and 6.67 °C
597 increase post-scour. During the years 1965-1967 in the Alsea Watershed Study maximum water
598 temperature increased 10 °C at the bottom of the Needle Branch Watershed following logging to
599 the stream's edge, stream cleaning, and slash burning (Ice, 2008). Gomi et al. (2006) found that
600 for four headwater streams subject to clearcut harvesting with no buffer retention, daily
601 maximum temperatures increased between 1.9 °C and 8.8 °C while increases for streams with
602 buffers ranged from 1.1 to 4.1° C, similar to our findings.

603
604 Gomi et al. (2006) reported that treatment effects were more subdued post-harvest with inclusion
605 of a 30 m buffer; maximum daily temperatures increased by <2 °C. When examining streams
606 with 10 m buffers, Jackson et al. (2001) found that two out of three streams produced an increase
607 of <2.4 °C and the third a change of -0.3 °C. Wilkerson et al. (2006) studied stream water
608 temperature response to harvesting in Maine with different buffer widths. Streams without
609 buffers experienced the largest temperature increases (1.4-4.4 °C). Stream with 11 m buffers
610 showed small, but not statistically significant increases (1.0-1.4 °C) while streams with 23 m
611 buffers as well as sites with partial harvest treatments showed no temperature increases. The
612 results from these contemporary studies including our findings for our private forest sites
613 indicate that some buffer retention practices likely reduce the magnitude of change but do not
614 necessarily completely eliminate harvest effects on stream temperature. However, temperature
615 increases and shade decreases were not ubiquitous among private sites; some did not indicate
616 increases in temperature following harvest (Figure 3). Other site conditions likely drove this
617 observed range of response.

618

619 4.3 Relationships between Shade and Riparian Characteristics

620 Increases in stream temperature were related to decreases in shade, both of which only occurred
621 on private sites. These results coincide with the findings of Groom et al. (2011). However, there
622 were ranges in shade and temperature responses on private sites reflecting variability in riparian
623 conditions after harvesting according to minimum Forest Practices Act regulations. The
624 variability may have been related to site differences in shade and factors related to shade.

625
626 Between 68 and 75% of variability in post-harvest shade may be accounted for by basal area
627 within 30 m of the stream, tree height, and possibly blowdown. Sites with higher basal area
628 within 30 m of the stream resulted in higher post-harvest shade. We anticipated that the variable
629 TPH would inform regression models containing basal area per hectare (BAPH), as total basal
630 area would depend on number of trees as well as tree size. The shade model ranking indicated
631 that inclusion of the TPH variable or its interaction with BAPH did not improve model fit; given
632 their correlation, we believe the two variables shared a similar relationship over this landscape.

633
634 Our findings suggest that sites with shorter trees had higher post-harvest shade. DeWalle (2010)
635 found in a modeling study that buffer height and density were as important as buffer width at
636 providing shade. However, their results predicted an increase of shade values with tree height
637 which is counter to our findings. It may be possible that our negative relationship between tree
638 height and shade is due to the negative correlation between crown ratios and tree height. Buffers
639 comprised of trees with canopies tens of meters above the stream may not protect streams from
640 mid-morning or afternoon sun exposure. We observed that private sites with low basal area
641 generally had taller trees. The relationship between the percent stream basal area represented by
642 hardwoods and height was not significant, suggesting that hardwood/conifer dominance did not
643 play a role. The negative relationship between plot blowdown basal area values and stream
644 shade appears reasonable, although the blowdown effect may be driven by extensive plot
645 blowdown at a single site.

646
647 We found that modeling stream shade as a function of riparian buffer width and the number of
648 stream banks harvested informative. While not as good a predictor as basal area, this model it
649 did explain 50% of the observed variability in post-harvest shade. Collectively the two models

650 suggest that harvesting timber within riparian areas on both sides of the stream, reducing basal
651 area and leaving narrower buffers are actions which contribute to decreases in shade.

652

653 5. Conclusion

654 Two years following timber harvest in our Oregon Coast Range streams, Maximum, Mean,
655 Minimum, and Diel Fluctuation summer temperatures increased along some sites. We detected
656 no differences between pre-harvest and post-harvest stream temperature on state forests,
657 indicating that state forest riparian buffers prevent harvest-related increases in shade and stream
658 temperature. Temperature increases on private sites were related to reduction in shade.
659 Reductions in shade were related to decreases in basal area for sites with greater tree heights.
660 Results correspond with the finding of elevated stream temperatures for private sites post-harvest
661 in the Groom et al. (2011) temperature standard analysis. Although our study's inference is
662 limited to the Oregon Coast Range, our determination of the relative efficacy of different buffer
663 designs at influencing shade and stream temperatures are likely relevant to other high-rainfall
664 low-order Douglas Fir dominated streams in the Pacific Northwest that are subject to similar
665 harvest practices.

666

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674

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686

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Table 1. Comparisons of different fixed-effects parameterizations for temperature models with similarly structured random effects. A set of model results are presented for each of the treatment reach changes in daily Maximum, Mean, Minimum, and Fluctuation values. Results are sorted by Maximum Δ AIC values. Rank indicates relative AIC order within a type of temperature analysis; a rank of 1 indicates the lowest AIC score. The symbol ω indicates model weight. All models include an intercept; linear combinations of independent variables are provided^a. See text for a definition of variables.

Model Name	Independent Variables ^a	Maximum			Mean			Minimum			Diel Fluctuation		
		Rank	Δ AIC	ω	Rank	Δ AIC	ω	Rank	Δ AIC	ω	Rank	Δ AIC	ω
Grad_Shade	GR, SH, TL, CT	1	0.00	0.40	1	0	0.43	2	0.4	0.20	3	0.7	0.18
ShadeGradWS	GR, SH, WS, TL, CT	2	0.40	0.33	2	0.8	0.29	5	1.6	0.11	4	1.7	0.11
StreamShade	SH, TL, CT	3	2.70	0.10	3	1.9	0.17	1	0	0.24	6	1.8	0.11
FullStream	GR, SH, WS, AZ, TL, CT	4	3.40	0.07	5	4.5	0.05	7	3.1	0.05	5	1.7	0.11
El_Shade	EL, SH, TL, CT	5	3.70	0.06	4	3.8	0.06	4	1.2	0.13	2	0.3	0.22
BO	GR, SH, WS, AZ, HS, OW, TL, CT, EL	6	5.30	0.03	6	7.5	0.01	12	7.1	0.01	1	0	0.26
Harvest_PXLength	PP, PP*TL, TL, CT	7	25.70	0.00	7	20	0.00	6	3	0.05	7	18.4	0.00
Harvest_Private	PP, TL, CT	8	33.60	0.00	8	23.1	0.00	3	1	0.15	8	29.4	0.00
Harvest	HS, TL, CT	9	57.50	0.00	9	38.6	0.00	10	6.5	0.01	9	49	0.00
Harvest_S_P	HS, OW, TL, CT	10	59.40	0.00	10	39.3	0.00	11	6.9	0.01	10	50.9	0.00
Upstream_TRlength	TL, CT	11	71.70	0.00	11	51.6	0.00	8	5.8	0.01	13	63.4	0.00
GradWS	GR, WS, TL, CT	12	72.30	0.00	12	52.8	0.00	14	8.1	0.00	17	66	0.00
Phys	GR, WS, EL, TL, CT	13	72.90	0.00	15	54.8	0.00	16	9	0.00	15	64.9	0.00
StreamAzimuth	AZ, TL, CT	14	73.60	0.00	13	52.9	0.00	9	6.3	0.01	16	65.4	0.00
Upstream	CT	15	73.90	0.00	17	56.3	0.00	17	15.6	0.00	12	61.9	0.00
AzimuthGradWS	AZ, GR, WS, TL, CT	16	74.00	0.00	14	53.7	0.00	15	8.2	0.00	18	68	0.00
El_Azimuth	EL, AZ, TL, CT	17	74.30	0.00	16	54.9	0.00	13	7.5	0.01	14	64.2	0.00
Intercept	--	18	74.50	0.00	18	61.5	0.00	18	16.3	0.00	11	60.5	0.00

^aGradient = GR, shade = SH, treatment length = TL, control temperature change = CT, watershed area = WS, valley azimuth = AZ, elevation = EL, owner = OW, Harvest Status = HS, private site post-harvest = PP, PP * treatment length = PPxTL.

Table 2. Annual weather and stream flow conditions at a representative weather station^a and stream gage^b. Air temperature data are the mean of daily maximum temperatures across the 40-day study period. Spring Precipitation is cumulative precipitation from 1 January to the beginning of the study period. Study Precipitation is cumulative precipitation during study period. Stream Flow is the average daily flow averaged over the study period.

Year	2002	2003	2004	2005	2006	2007	2008
Air Temperature (°C)	24.4	26.0	25.8	26.5	25.1	21.8	24.2
Spring Precipitation (cm)	126.9	134.5	139.9	100.7	129.8	93.9	112.5
Study Precipitation (cm)	0.38	0.30	1.55	0.30	0.08	5.56	6.15
Stream Flow (L/s)	53.1	44.3	72.5	79.7	51.4	52.5	72.8

^aRye Mountain Remote Area Weather Station. Lat: 45.2172; long: -123.5358.

^bU.S. Geological Survey gage 14303200 near Blaine, OR. Lat: 45.1928; long: 123.3243.

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Table 3. AIC ranking for seven of the eight Shade models. Only the R^2 value is presented for the model Blowdown as the parameter BD_Channel had 31 observations. For all other variables $n = 33$.

Model	Variables	k	$\Delta AICc$	ω	R^2
BasalXHeight	BAPH, Height, BAPH*Height	4	0.00	0.99	0.69
BasalHeight	BAPH, Height	3	10.70	0.00	0.50
BasalCR	BAPH, CR	2	15.08	0.00	0.43
BasalArea	BAPH	4	16.10	0.00	0.33
BufferWidth		4	16.91	0.00	0.33
	NSides, BuffWidth, Nsides * BuffWidth				
TreesXBasal	TPH, BAPH, TPH*BAPH	3	17.89	0.00	0.46
Intercept	--	1	25.03	0.00	0.00
Blowdown	PlotBD, ChannelBD	--	--	--	0.06

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Table 4. Fixed- and random-effect parameter values for four linear mixed-effects models and their associated temperature response variables. Model StreamShade lacks a parameter for gradient and model El_Shade replaces gradient with the parameter elevation. Parameters for treatment length and gradient are expressed as change in temperature per 1 km of distance or elevation. Observations = 119, Groups (Sites) = 33.

Response	Maximum Temperature				Mean Temperature				Minimum Temperature				Diel Fluctuation ^b			
Model	Grad_Shade				Grad_Shade				StreamShade				El_Shade			
Fixed ^a	DF	Value	SE	p	DF	Value	SE	p	DF	Value	SE	p	DF	Value	SE	p
Intercept	29.1	0.494	0.125	0.001	28.1	0.282	0.080	0.001	28.6	0.183	0.059	0.004	30.2	0.238	0.094	0.017
CT	21.5	-1.232	0.459	0.014	21.7	-1.230	0.345	0.002	22.8	-0.518	0.310	0.109	23.2	-0.783	0.583	0.193
TL	28.2	0.800	0.304	0.014	27.6	0.638	0.194	0.003	27.8	0.549	0.144	0.001	31.5	0.627	0.279	0.031
SH	94.5	-5.866	0.572	0.000	97.5	-3.050	0.371	0.000	101.0	-0.881	0.314	0.006	99.6	-4.698	0.503	0.000
GR ^b	30.3	-0.076	0.036	0.040	29.6	-0.043	0.023	0.067	--	--	--	--	31.1	-0.104	0.057	0.079
Random	Std.Dev				Std.Dev				Std.Dev				Std.Dev			
Intercept	0.441				0.181				0.095				0.241			
ControlTemp	3.564				1.134				0.975				6.175			
Residual	0.079				0.040				0.031				0.060			

^aControl reach temperature change = CT, gradient = GR, shade = SH, treatment length = TL.

^bFor Diel Fluctuation the variable for GR is replaced by elevation (EL). Other parameters in the model are the same.

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Table 5. Mean and range values for State and Private independent variables and site characteristics. Values are calculated from 15 State sites and 18 Private sites. Pre and Post refer to measurements taken preharvest or postharvest. For Shade ranges see Figure 4; basal area and trees per hectare are BAPH and TPH, respectively.

Variable	State		Private	
	Mean	Range	Mean	Range
Gradient (%)	6.5	1.5 - 13.2	6.4	1.0 - 17.5
Treatment Length (km)	0.8	0.3 - 1.5	0.6	0.3 - 1.8
Elevation (m)	350	160 - 570	300	3 - 900
Watershed Area (ha)	222	72 - 593	208	27 - 626
Crown Ratio	0.43	0.30 - 0.56	0.40	0.26 - 0.57
Buffer Width (m) ^a	51.8	25 - 61	31	19 - 41
Bankfull Width (m)	4.6	2.7 - 7.9	4.1	2.2 - 7
Wetted Width (m)	2.3	1.3 - 3.7	2.0	1.0 - 3.0
Thalweg (cm)	17	9 - 30	15	8 - 24
Basal Area (m ² /ha)				
Pre-harvest	41	19 - 74	43	23 - 73
Post-harvest	42	25 - 73	25	11 - 40
Trees per Ha				
TPH Pre	368	147 - 665	465	196 - 664
TPH Post	387	128 - 645	270	111 - 429
Tree Height (m)	26	17 - 37	25	18 - 31

^aMeans reported in Groom et al. (2011); 95% CI for State sites = 45.6 m, 58.0 m; 95% CI for Private sites = 26.7 m, 35.3 m.

Table 6. Fixed- and random-effect parameter values for the Harvest_Private model fit by restricted maximum likelihood. Observations = 119, Groups (Sites) = 33. TRLength is expressed as change in temperature over 1 km.

Parameters	<u>Maximum Temperature</u>				<u>Mean Temperature</u>				<u>Minimum Temperature</u>				<u>Diel Fluctuation</u>			
	DF	Value	SE	<i>p</i>	DF	Value	SE	<i>p</i>	DF	Value	SE	<i>p</i>	DF	Value	SE	<i>p</i>
Fixed																
Intercept	33.9	0.280	0.143	0.058	32.7	0.191	0.088	0.037	32.3	0.152	0.060	0.017	33.1	0.076	0.116	0.515
CT	21.5	-0.798	0.520	0.140	18.8	-1.018	0.393	0.018	24.2	-0.407	0.315	0.209	23.4	-0.528	0.631	0.412
TL	30.2	0.745	0.335	0.034	29.4	0.602	0.208	0.007	27.8	0.538	0.141	0.001	30.1	0.216	0.275	0.440
PP	92	0.711	0.101	0.000	92.4	0.369	0.063	0.000	91	0.128	0.049	0.011	88.6	0.578	0.084	0.000
Random			Std.Dev				Std.Dev				Std.Dev				Std.Dev	
Intercept			0.535				0.206				0.091				0.355	
CT			4.554				1.596				1.036				7.008	
Residual			0.111				0.049				0.032				0.076	

^aControl temperature change = CT, treatment length = TL, private sites post-harvest = PP.

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Figure Captions

Figure 1. Schematic of a stream reach (from Dent et al. 2008). Data loggers were placed starting upstream at Stations 1, 2, and 3. The upstream Control reach laid between Stations 1 and 2, the treatment reach between Stations 2 and 3. Riparian Data Collection Plots were located on either side of the stream in the Treatment and Control reaches; in this study we only used riparian data from the Treatment reach plots.

Figure 2. Predicted temperature response for Maximum, Mean, Minimum, and Diel Fluctuation metrics at the observed extreme values of single explanatory variables and at the mean of all other variables. Explanatory variables presented appear in the models Grad_Shade and El_Shade. The El_Shade model, best supported by the Diel Fluctuation temperature metric, differs from Grad_Shade in that it replaces the Gradient parameter with Elevation. Influences of Gradient and Elevation extremes for those models appear in the lower right figure. Circles in each figure represent point estimates with 95% confidence intervals. In the upper left graph control temperature change represents change over 100 m.

Figure 3. Observed and predicted temperature changes for Maximum Temperatures (°C) by site. Pre-harvest and post-harvest observations are represented by open and filled circles, respectively. Each point represents one year of data collection at a site. The crosses represent predicted values from model Grad_Shade. Above each site's data is listed its site number, ownership ([S]tate or [P]rivate), post-harvest shade value, and in parentheses the change in shade value pre-harvest to post-harvest. Sites are ordered from the upper left to lower right by the observed change in shade values. Vertical differences of points within a pre-harvest or post-harvest category indicate a between-year change in the temperature relationship between Stations 1 and 2.

Figure 4. Plot of average treatment reach shade values (%) for each site, grouped by harvest status (pre-harvest, post-harvest). On the left are state forest shade values, on the right shade are private forest values.

Figure 5. Post-harvest values for treatment reach (A) logit-transformed percentages of mean shade values, (B) tree heights, and (C) basal area (m^2/Ha) by Treatment Reach length. Triangles

represent privately-owned sites; circles are state sites. The dashed line represents the best linear fit for the filled circles ($y\text{-axis variable} = \text{intercept} + x\text{-axis variable}$). Shade values ranged from 0.51 to 0.95.

Figure 6. Within-site pairwise differences in temperature change between post-harvest and pre-harvest values for Maximum observed data and partial residuals. Observed values are presented individually in Figure 3. Partial-residual values represent observed values but control for site treatment reach length, upstream control temperature change, and stream gradient.