

Natural and Anthropogenic Controls on Wood Loads in River Corridors of the Rocky, Cascade, and Olympic Mountains, USA

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This is a peer-reviewed manuscript published in *Water Resources Research*. Note that due to copyediting, there may be slight differences between this version and the published version, but the content is almost entirely identical. The final, published version of this manuscript is available via the published DOI on this page.

Abstract

Wood in rivers creates habitat, shapes the morphology of valley bottoms, and acts as a pool of organic carbon (OC). Effective riverine wood management depends on a robust understanding of the spatial distribution of wood throughout river networks. This motivates the analysis of wood load in relation to both reach- and basin-scale processes. We present wood load data coupled with precipitation, forest stand characteristic, land use, and geomorphic data across four basins in the Rocky, Cascade, and Olympic Mountains of the western USA. We compare basins with differing land use within the same climatic region and basins in differing climates and statistically model intra-basin wood load variability. Wood load is a function of metrics that generally describe river corridor spatial heterogeneity, metrics that describe wood storage patterns, and, at a broader scale, metrics that relate to wood supply. From this, we generate a conceptual model to describe controls on wood load across spatial scales. We use this model to propose that spatial heterogeneity and wood storage pattern together determine reach-scale wood trapping efficiency. Trapping efficiency in turn regulates how wood supply to valley bottoms determines wood load. We also find that wood in an undisturbed basin stores significant amounts of OC, and that wood load restoration has the potential to restore significant amounts of OC to valley bottoms. This conceptual model of wood load controls may serve as a framework to guide wood load modeling and restoration at multiple scales.

Key Points

- Wood load in 4 basins is controlled by wood supply, spatial heterogeneity, and storage pattern
- A paired basin study reveals logging to have reduced wood loads by a factor of two
- Wood acts as a significant OC storage mechanism in valley bottoms, especially in undisturbed systems

Plain Language Summary

Downed wood in rivers creates habitat and nutrients for organisms in streams and on floodplains. Humans have negatively impacted valley bottoms through the removal of downed wood. We measured the amount of downed wood in valley bottoms in four mountain river basins to understand what factors, both local and regional, determine how much wood is stored in river

44 corridors. We found that at the regional scale, logging, precipitation, and forest characteristics
45 control the supply of wood to valley bottoms. At a more local scale, the shape of the valley
46 bottom and the way in which wood is stored (either as accumulations known as jams or as
47 individual logs) determine how much wood can be trapped in the valley bottom. We present a
48 conceptual model that ties these factors together and can guide our understanding and
49 management of how much wood is in rivers.

50 **1 Introduction**

51 Wood accumulates in rivers via bank erosion and mass movements from hillsides.
52 Because wood can remain stable in the channel and on floodplains, it plays a foundational role in
53 shaping the ecology and geomorphology of valley bottoms. By providing colonization surfaces
54 for periphyton and macroinvertebrates as well as a source of carbon, wood increases
55 microhabitat diversity and provides energy input to macroinvertebrates (Benke & Wallace, 2003;
56 Wondzell & Bisson, 2003). By shaping the location, abundance, and geometry of pools and
57 altering bed texture (Gomi et al., 2003; Montgomery et al., 1996, 2003), wood can regulate
58 habitat abundance and diversity for fishes (Jones et al., 2014; Nagayama et al., 2012). Wood also
59 serves as a pool of organic material (Naiman et al., 1987; Osei et al., 2015; Sutfin et al., 2016),
60 providing a source of organic matter as it breaks down and impacting terrestrial organic carbon
61 (OC) cycling (Elosegi et al., 2007; Wohl, Hall, et al., 2017). The potentially substantial role of
62 wood in storing OC motivates quantification of wood loads not only in terms of wood volume
63 per unit area, but also as an estimated OC stock (mass per unit area).

64 Wood loads are a function of channel geometry, land use, bioclimatic regime, and
65 geomorphic processes (Wohl, Lininger, et al., 2017). Unit wood volume in valley bottoms
66 sometimes correlates inversely with channel width and drainage area (Beechie & Sibley, 1997;
67 Bilby & Ward, 1989, 1991; Wohl, Lininger, et al., 2017), although this correlation has been
68 observed to be direct in some cases and is strongly dependent on bioclimatic region and riparian
69 forest characteristics (Burton et al., 2016; Wohl, Lininger, et al., 2017). Wood loads have little
70 consistent relation to channel characteristics across bioclimatic regions, but individual regions
71 and watersheds do display significant trends, allowing wood load to be predicted by variables
72 describing geomorphic, ecologic, and anthropogenic conditions (Hough-Snee et al., 2015; Wohl,
73 Lininger, et al., 2017). While mechanisms influencing wood transport and storage have been
74 explored in flume environments (Bocchiola et al., 2006; Braudrick et al., 1997; Davidson et al.,
75 2015), we still lack a good understanding of how well experimental results translate to natural
76 conditions. Such a mechanistic understanding is necessary to explain differences between
77 bioclimatic regions (e.g., why wood load and drainage area sometimes correlate directly and
78 sometimes inversely) and explain wood load spatial distribution across scales. This motivates us
79 to seek a mechanistic understanding of the controls on wood load across spatial scales.

80 By regulating storage pattern and mobility, wood jams are a potential mechanistic control
81 on wood transport and wood load. Wood jams are generally more stable than dispersed wood
82 pieces within a given reach (Dixon & Sear, 2014; Ruiz-Villanueva et al., 2016; Wohl & Goode,
83 2008). However, wood jams are not uniformly distributed throughout river networks (Benda,
84 1990; Cadol et al., 2009; Marcus et al., 2002; Pfeiffer & Wohl, 2018). This implies that the
85 importance of wood jams and their impacts on wood transport may vary with network position,
86 likely due to differences in piece mobility, which is strongly regulated by stream size relative to
87 the length of wood pieces (Gurnell et al., 2002; Kramer & Wohl, 2016). This motivates an
88 analysis of the importance of jams in regulating wood loads relative to other variables that might
89 impact wood loads.

90 Forest management, especially in the form of timber harvest, is one of the most
91 widespread human impacts on forests in mountainous regions. Logging commonly impacts wood
92 in valley bottoms by influencing both riparian recruitment rates and the rate at which mass
93 movements transfer wood from hillslopes to rivers. Logging and associated road-building
94 increase the rate of mass wasting on steep slopes (Guthrie, 2002; Jakob, 2000; Roberts et al.,
95 2004; Sidle et al., 2006; Wolter et al., 2010), potentially increasing the delivery of wood to
96 floodplains and channels. However, the widespread wood removal and streamside harvesting of
97 wood associated with clearcutting in many regions has a net effect of reducing wood loads and
98 reducing wood trapping ability by reducing in-stream roughness and spatial heterogeneity (Hyatt
99 & Naiman, 2001; Ruffing et al., 2015; Wohl, 2014). This loss of roughness can act as a negative
100 feedback on wood storage, leading to high rates of wood export from the system even after
101 riparian corridors have reforested (Bilby, 1984). The impact of logging on in-stream wood has
102 been demonstrated dominantly through the loss of large wood pieces (Bilby & Ward, 1991;
103 Ralph et al., 1994), which could reduce the occurrence of relatively stable wood jams. Past
104 examinations of the effects of logging have commonly used data from individual reaches in a
105 variety of basins with differing conditions (and potentially confounding variables). This
106 motivates a more rigorous examination of the effects of logging across the entirety of river
107 networks.

108 1.1 Objectives

109 Here, we seek to move towards a mechanistic and multi-scale understanding of the
110 controls on wood loads by quantifying wood loads across a diverse set of mountain river basins
111 and modeling relationships between those wood loads and the natural and anthropogenic
112 (namely, logging) processes that impact them. To our knowledge, we provide the first field-
113 based quantification of wood load across the entirety of our four study basins, allowing a
114 rigorous examination of the intra-basin trends in wood load from headwaters to basin outlet. By
115 quantifying wood loads and using published data on wood density and OC content, we also seek
116 to apply our examination of wood load variability to variability in wood OC storage.

117 Previous broad-scale studies of wood load spatial variability (Hough-Snee et al., 2015;
118 Wohl, Lininger, et al., 2017) generally conclude that wood loads can be conceptualized at either
119 broad (inter-basin) or local (intra-basin) scales by taking into account either bioclimatic or site-
120 specific variables (e.g., land use, channel geometry), respectively. However, a conceptual model
121 to describe wood load spatial distribution that applies at all scales has yet to be developed. We
122 use our extensive field dataset and statistical analyses to suggest that a single conceptual
123 framework can be used to guide understanding of wood load spatial variability both within
124 (intra-basin) and between (inter-basin) river basins.

125 We use statistical modeling of field-sampled wood load data from four mountain river
126 basins in three distinct regions across the western USA to determine the dominant controls on
127 wood load both within each basin (intra-basin) and between basins (inter-basins). By considering
128 wood supply and mechanistic variables relating to reach-scale wood trapping efficiency, we
129 develop a novel conceptual understanding of wood load spatial variability that applies to
130 multiple scales. This conceptual model explains our results and provides a basis for further
131 testing of multi-scale controls on wood load spatial distribution in river networks.

132 2 Methods

133 2.1 Field Sites

134 Our choice of study basins maximizes variability within the western United States in
135 factors that may influence wood loads (forest stand characteristics, valley morphology, climate,
136 etc.), allowing for a robust analysis of wood load spatial variability. We quantified basin-scale
137 wood load in the Big Sandy basin in the Wind River Range of Wyoming, the Middle Fork (MF)
138 Snoqualmie basin in the central Cascade Mountains of Washington, and the Sitkum and South
139 Fork (SF) Calawah River basins in the Olympic Mountains of Washington (Figure 1). These
140 basins represent three distinct bioclimatic and geomorphologic regions, ranging from the semi-
141 arid Middle Rockies to the wet, glacially influenced Cascades and more fluvially dominated
142 basins in the Olympics. Mean annual precipitation, relief, drainage area, and mean basin slope
143 for each study basin are given in Table 1.

144 We performed a paired basin study using the Sitkum and SF Calawah basins to examine
145 the effects of basin-wide clearcut timber harvest. These two basins are of similar network
146 geometry (Figure 1) and are both underlain by marine sedimentary rocks (Gerstel & Lingley Jr.,
147 2000). Forests in both basins are dominated by Douglas fir (*Pseudotsuga menziesii*), Sitka spruce
148 (*Picea sitchensis*) and western hemlock (*Tsuga heterophylla*). The SF Calawah lies entirely
149 within the boundary of Olympic National Park and has not experienced forest harvest or road
150 building, in contrast to the Sitkum, which has been clearcut extensively since the 1940s (orange
151 overlay in Figure 1). Road building and clearcut timber harvest were widespread in the Sitkum
152 until the 1990s, with 15 m (50 ft) riparian buffers being implemented on many reaches in 1975.
153 Currently, forests are dominantly being thinned and roads are being decommissioned to enhance
154 forest habitat and reduce mass movement frequency. The result of this land use has been the loss
155 of large trees able to be recruited to streams by bank and hillslope failure (Pacific District
156 Olympic National Forest, 2012).

157 We also present data from the MF Snoqualmie and Big Sandy Rivers to examine a
158 bioclimatic contrast that allows us to examine wood loads in regions with differing precipitation,
159 forest characteristics, and network structure. The MF Snoqualmie exhibits glaciogenic
160 topography, with streams ranging from steep, debris flow dominated headwater channels to
161 lower gradient, wide, laterally unconfined channels in its lower reaches, and has been extensively
162 logged in its lower elevation reaches. The elevation range in the MF Snoqualmie generates a
163 strong vegetation gradient. The talus, active glaciers, and alpine tundra at the highest elevations
164 grade to subalpine forests dominated by mountain hemlock (*Tsuga mertensiana*) (above
165 approximately 1500 m), but also including Pacific silver fir (*Abies amabilis*) and noble fir (*Abies*
166 *procera*) grading into the montane zone (above approximately 900 m). At lower elevations,
167 uplands and terraces are covered by Douglas fir (*Pseudotsuga menziesii*) and western hemlock
168 (*Tsuga heterophylla*), whereas active riparian zones are dominated by red alder (*Alnus rubra*)
169 and bigleaf maple (*Acer macrophyllum*).

170 The Big Sandy also exhibits glaciogenic topography, but is much drier than the MF
171 Snoqualmie. While higher elevations (above approximately 3100 m) are characterized by
172 herbaceous alpine tundra, the subalpine zone (approximately 2900 to 3100 m) is characterized by
173 forests of whitebark pine (*Pinus albicaulis*), Engelmann spruce (*Picea engelmannii*), and
174 subalpine fir (*Abies lasiocarpa*). The montane zone (approximately 2600 m to 2900 m) is
175 comprised dominantly of lodgepole pine (*Pinus contorta*). Only a small portion of this basin
176 (approximately 1%) resides below 2500 m, where shrub steppe begins to dominate (Fall, 1994).
177 Forests in this basin are patchy, with substantial grassy parklands and meadows.

178 To simplify our presentation of results and highlight contrasts between these basins, we
179 categorize these basins by land use, climate, and geomorphic legacy. We term the MF
180 Snoqualmie, with its wet climate, and glacially carved lakes and broad valley bottoms as the *wet*
181 *glaciogenic* basin. In contrast, we term the Big Sandy, with its semi-arid climate, and broad,
182 glacially carved valley bottoms as the *semi-arid glaciogenic* basin. Finally, we term the Sitkum
183 and SF Calawah, which exhibit the wettest climate, but most fluvially incised, narrow valley
184 bottoms as the *wet fluviogenic* basins. We further subset the Sitkum as the *logged* wet
185 fluviogenic basin and the SF Calawah as the *unlogged* wet fluviogenic basin (Figure 1).
186

187 2.2 Study Design and Sampling

188 We sampled basins in summer 2016 (both wet fluviogenic basins and the semi-arid
189 glaciogenic basin) and summer 2017 (wet fluviogenic basin). Sampling during the summer
190 ensured that there were no large, wood-transporting floods during sampling, such that our data
191 represent an estimate of the wood load in each basin at a single time. We collected a total of 148
192 reach-scale (each reach is 100 m or 10 channel widths long, whichever was shorter) samples of
193 valley bottom wood load across all four study basins.

194 We used stratified random sampling to generate an unbiased sample of wood load
195 measurement sites in each basin. In the semi-arid glaciogenic basin, we used a combination of a
196 10 m DEM and satellite imagery to manually map the extent of the valley bottom along the
197 entire stream network, with the objective of delineating confined and unconfined valley sections.
198 We defined unconfined valley bottoms as those in which channel width occupied no more than
199 half the valley bottom, and confined valley bottoms as those in which channel width occupied
200 greater than half the valley bottom. We then stratified the stream network by five drainage area
201 classes to ensure uniform sampling throughout the basin. This produced two stratifications, one
202 of drainage area and the other of confinement. In the wet basins, the dense vegetation prevented
203 us from manually mapping valley bottoms as we did in the semi-arid basin. Thus, in the wet
204 fluviogenic basins, we sampled uniformly across stream orders (Strahler, 1957) in order to
205 sample a relatively even distribution of channel and valley widths. We stratified the wet
206 glaciogenic basin stream network by slope into four strata. We chose to not measure wood loads
207 in parts of the network steeper than 0.30 m/m as classified by a 10 m DEM because our initial
208 field reconnaissance indicated that many such channels were dominated by colluvial processes as
209 opposed to fluvial processes, although field-based measurements indicated that some study sites
210 were locally steeper than this threshold. Within each slope strata, we randomly selected ten
211 reaches for sampling wood load.

212 In all four basins, but especially in the wet fluviogenic basins, we were unable to reach all
213 randomly sampled sites due to time constraints. This resulted in the subjective selection of sites
214 that were accessible and that we felt maintained as unbiased a sample as possible. Total numbers
215 of sites and the proportion of sites that were subjectively chosen are listed in Table S1.

216 2.3 Reach-Scale Field Measurements

217 Table S1 summarizes which measurements were collected in each basin. Within each
218 reach, we quantified wood volume in wood jams (accumulations of 3 or more pieces touching
219 one another) using a census approach, measuring the length, width, and height of a rectangular
220 prism that best fit the jam (i.e., these geometric measurements did not correspond to flow
221 direction) and visually estimating the porosity (Thevenet et al., 1998). Although this method is
222 not as accurate as dismantling jams to measure every wood piece (e.g., Manners et al., 2007), our

223 consistency in this method (i.e., only a single person made all estimates using consistent
224 methodology) likely minimizes systematic bias. Within each reach, we quantified wood volume
225 in dispersed pieces greater than 10 cm diameter using a combination of two methods, depending
226 on the nature of wood within the reach and channel confinement. For confined valleys with
227 numerous wood pieces dominantly oriented perpendicular to the valley axis, we used an adapted
228 form of a line-intersect sampling strategy (Van Wagner, 1968; Wallace & Benke, 1984) whereby
229 the line was fit to the channel centerline (Warren et al., 2008). We measured the diameter of
230 every wood piece intersected by the line, then calculated wood volume using the formula given
231 by Van Wagner (1968). For unconfined reaches with sufficiently low wood piece abundance, we
232 measured the diameter and length of each wood piece in the reach, calculating piece volume as if
233 each piece was a cylinder. In the wet glaciogenic basin, some unconfined floodplains were wide
234 enough that a census of pieces and jams was impractical, so we performed a census within the
235 channel, then performed a single line intersect transect across the floodplain perpendicular to the
236 valley axis to quantify floodplain wood load (Van Wagner, 1968).

237 We assigned a decay class to each reach that describes all the pieces and jams in each
238 reach using the visual decay classification of Harmon et al. (2011). This allowed us to estimate
239 an average wood density using the downed dead softwood densities for each decay class listed in
240 Table 5 of Harmon et al. (2011). With an average wood density and volume per reach, we
241 calculated wood mass as the product of density and volume. We used the length of each reach
242 and the valley bottom width to compute a wood mass per unit area of valley bottom.

243 At each reach, we measured channel geometry and other characteristics using a TruPulse
244 360 laser rangefinder (Scott et al., 2016), although our measurements were not consistent across
245 all basins because field protocol evolved during the course of the study (Table S1). In the wet
246 glaciogenic basin, we categorized channels by planform and dominant bedform (Montgomery &
247 Buffington, 1997). We defined planforms as either: straight, where the channel was generally
248 confined and significant lateral migration was not evident; meandering, where lateral migration
249 was evident but only a single channel existed; anastomosing, where vegetated islands separate
250 multiple channels; and anabranching, where a single dominant channel existed with relict
251 channels separated by vegetated islands. For the purposes of statistical modeling, we also
252 classified channels as being either multithread (anastomosing or anabranching) or single thread
253 (straight or meandering). Because logging records are inconsistent and likely inaccurate in the
254 wet glaciogenic basin (based on the frequent observation of past logging activity where none was
255 recorded in Forest Service records), we noted whether signs of logging, such as cut stumps,
256 cable, decommissioned roads or railroads, or other logging-associated tools were found near the
257 reach. We also looked for forest stand characteristics that commonly result from clearcut
258 logging: even-aged stands, monocultures, and a lack of undergrowth compared to unlogged
259 forests. These observations, and our resulting classification of reaches as being logged or
260 unlogged, are limited to the forests immediately surrounding the reach.

261 2.4 GIS and Derivative Measurements

262 A 10 m DEM was utilized for all topographic measurements. We collected the following
263 data for each reach using a GIS platform: elevation, drainage area, land cover classification and
264 canopy cover from the National Land Cover Database (Homer et al., 2015), and the mean slope
265 of the basin upstream of each reach. Utilizing drainage area at each reach and field-measured
266 slope, we calculated an estimated stream power as the product of drainage area, slope, and basin-
267 averaged precipitation.

268 We calculated a wood jam density to measure the abundance of wood jams in each reach
269 as the number of jams divided by the length of each reach. Following Kramer and Wohl (2016),
270 we calculated a dimensionless maximum piece length for each reach (L^*) as the maximum piece
271 length in the reach divided by the bankfull width, for all reaches except those in the semi-arid
272 glaciogenic basin, where bankfull width was not measured. All wood masses were normalized by
273 unit area using the average valley width and length of each reach. For purposes of estimating OC
274 storage in wood, we assumed that half of the measured wood mass was carbon (Lamlom &
275 Savidge, 2003). Variability in wood OC content ranges from 47.21% to 55.2% for conifers
276 (Lamlom & Savidge, 2003), the dominant division of trees present in our study basins. As such,
277 an assumption of 50% OC content is likely a conservative estimate of actual OC content and is a
278 suitable approximation for making first-order estimations of wood OC stock (e.g., Sutfin et al.,
279 2016; Wohl et al., 2012).

280 2.5 Statistical Analyses

281 All statistical analyses were performed using the R statistical computing software (R
282 Core Team, 2017). Due to differences in variables measured for each region, we conducted
283 modeling based on model groups with consistent measurements. We modeled wood load in the
284 wet glaciogenic (sample size, $n = 46$) and semi-arid glaciogenic ($n = 52$) basins individually as
285 well as across both fluvio-genic basins combined ($n = 50$). Because of the lack of variation in
286 hypothesized predictor variables in other basins, we only modeled the proportion of wood in
287 jams in the wet glaciogenic basin. We note that although this modeling predicts wood load as a
288 mass per unit area, we observe a Pearson correlation coefficient with a 95% confidence interval
289 between 0.98 and 0.99 between wood mass per unit area and wood volume per unit area. We also
290 tested each final model using wood volume as a predictor to ensure that results reported here are
291 equally applicable to wood volume and wood mass.

292 Our modeling strategy starts with univariate analysis between each hypothesized
293 predictor and the response, utilizing mainly comparative Wilcoxon rank-sum tests (Wilcoxon,
294 1945) or Spearman correlation coefficient statistics. During this filtering, we also view boxplots
295 or scatterplots as appropriate to discern which variables appear to have anything other than a
296 completely random relationship with the response. We then utilize all subsets multiple linear (for
297 wood load) or multiple logistic (for the proportion of wood stored in jams) regression using the
298 corrected Akaike Information Criterion as a model selection criteria (Wagenmakers & Farrell,
299 2004). We iteratively transform response variables to ensure homoscedasticity of error terms.
300 When selecting a single best model, we utilize both Akaike weight based importance as well as
301 parsimony to select a final, reduced model. We consider sample sizes, p values, and effect
302 magnitudes (odds ratios for logistic regression and slope coefficients for linear regression) in our
303 discussion of variable importance. All other statistical analyses presented here are comparative
304 statistics utilizing Wilcoxon rank-sum tests or pairwise equivalent using a Holm multiple-
305 comparison correction (Holm, 1979) to accommodate generally skewed distributions. Unless
306 otherwise noted, we present 95% confidence intervals to represent variance on population
307 estimates.

308 3 Results

309 3.1 Controls on Wood Load

310 Median wood load is significantly different between all study basins except for the wet
311 glaciogenic and logged wet fluvio-genic basins (Figure 2, Table S2). Wood load is highest in the

312 unlogged wet fluvio-genic basin, followed by the wet glaciogenic and logged wet fluvio-genic
313 basins, followed by the semi-arid glaciogenic basin. Although not shown here, trends in wood
314 volume between basins track very similarly to those in wood mass (Figure S1). Distributions of
315 wood loads are generally right skewed, especially in the fluvio-genic basins. For the entire dataset
316 of wood load in each basin, see Dataset S1 (Scott & Wohl, 2018). Table S3 shows the variables
317 tested to understand controls on wood load and a summary of results for each model.

318 Multiple linear regression modeling of wood mass per unit area in the wet glaciogenic
319 study basin reveals jam density (number of jams per meter), elevation, estimated stream power,
320 and confinement to be significant controls on wood load (adjusted $R^2 = 0.40$, $p < 0.0001$). For
321 this model, a cube root transformation (to accommodate 0 values) is found to be appropriate, so
322 all slope coefficients relate to a unit increase in the cube root of wood load. We note that the
323 cube root of wood load, while uninterpretable in itself, is likely analogous to a wood length per
324 unit area, if mass and volume are taken to be highly correlated (which they are in our data, see
325 section 2.5) A higher jam density (units of jams/m stream, $\beta = 9.04 \pm 8.16$) and higher estimated
326 stream power (units of m^3 , $\beta = 2.13 \times 10^{-8} \pm 2.06 \times 10^{-8}$) result in higher wood loads, whereas
327 higher elevations result in lower wood loads (units of m, $\beta = -1.20 \times 10^{-3} \pm 7.38 \times 10^{-4}$).
328 Unconfined streams are found to generally store less wood ($\beta = -.48 \pm 0.44$); all other predictors
329 held constant. We note that the effect (β) of stream power on wood load is extremely small,
330 despite its significance in the model. From this, we conclude that although stream power likely
331 has some relation to wood load, its effect is so much smaller than other controls that it is
332 negligible.

333 Similarly, in the semi-arid glaciogenic basin, jam density, elevation, and confinement in
334 addition to median piece length are found to be significant predictors of wood load (adjusted R^2
335 $= 0.77$, $p < 0.0001$). However, we find that piece length and confinement were strongly related,
336 leading to multicollinearity in any model including both variables. Comparing models similar to
337 the above model but with either confinement (adjusted $R^2 = 0.59$, $p < 0.0001$) or piece length
338 (adjusted $R^2 = 0.71$, $p < 0.0001$) removed, the model that includes confinement explains much
339 more of the variance in wood load. As such, we conclude that confinement is likely the dominant
340 control on wood load over piece length, and eliminate piece length from the final model. Thus,
341 our final model of wood load in the semi-arid glaciogenic basin includes only jam density,
342 elevation, and confinement as significant predictors of the cube root of wood load. Reaches with
343 higher jam densities (units of jams/m, $\beta = 14.38 \pm 6.94$) and lower elevations tend to store more
344 wood (units of m, $\beta = -0.0012 \pm 0.00048$). Like the wet glaciogenic basin, unconfined reaches
345 store significantly less wood than confined reaches ($\beta = -0.74 \pm 0.20$).

346 The logged wet fluvio-genic basin contains half as much wood as the unlogged wet
347 fluvio-genic basin (section 3.1.1). After accounting for logging, channel slope and jam density are
348 significant predictors of the cube root of wood load (adjusted $R^2 = 0.34$, $p < 0.0001$) in these
349 basins. Reaches with higher channel slope (units of m/m, $\beta = 2.66 \pm 0.62$) and more jams tend to
350 store more wood (units of jams/m, $\beta = 19.13 \pm 1.63$).

351 In summary, we find that jam density, elevation, and confinement in the glaciogenic
352 basins; and logging, slope, and jam density in the fluvio-genic basins control wood load. We
353 broadly categorize these variables into those that describe wood supply to valley bottoms
354 (elevation and logging) and those that describe reach-scale wood trapping efficiency (jam
355 density, confinement, and slope).

356

3.1.1 Effects of Logging on Wood Loads

357

358 Comparing the logged wet fluvio-genic (extensively clearcut) to the unlogged wet
359 fluvio-genic basin (relatively pristine), we find that wood loads are a factor of 2 greater in the
360 unlogged basin (Figure 2, Table S2). Other variables such as bankfull width, slope, wood jam
361 density per unit stream length, median and maximum piece length and diameter do not
362 significantly differ between basins (p values for comparisons are 0.70, 0.24, 0.47, 0.26, 0.19,
363 0.43, 0.70, respectively). We do note maximum piece diameter may be lower in the logged wet
364 fluvio-genic basin, and that we may lack the sample size to note this effect. The only factor that is
365 significantly different between basins is elevation, which is significantly higher in the logged wet
366 fluvio-genic basin ($p < 0.001$). However, we note that elevation was not found, either through
367 univariate analysis ($p = 0.56$) or model selection, to be a meaningful predictor of wood load
368 when modeling controls on wood load across samples in both fluvio-genic basins, likely due to
369 the lack of variation in forest stand characteristics with elevation in these basins.

369

370 Because historic logging records are largely inaccurate in the wet glaciogenic basin, we
371 use our observational mapping of logging to understand logging extent and attempt to understand
372 how variation in logging impacted wood load. We find that with very few exceptions, all sites at
373 low elevations experienced some form of timber harvest, likely within the last century. When
374 considering all sampled reaches in the basin in a univariate analysis, we find that sites with
375 logging apparently contain more wood than sites with no logging nearby ($p = 0.04$). However,
376 we also find that logging is strongly correlated with elevation, such that the median elevation of
377 logged sites (446_{-119}^{+71} m) is less than half that of unlogged sites (989_{-66}^{+173} m). Elevation is a
378 significant predictor of wood load in this basin due to the high range of elevation and forest
379 types. This suggests that the correlation between local logging activity at a reach and enhanced
380 wood loads in this basin is spurious, and that local impacts of logging cannot be evaluated here.

380

381 Smithwick et al. (2002) measured potential carbon stores in forests of the Pacific
382 Northwest, including the Washington Cascades and Olympic Mountains. We utilize
383 measurements of downed log OC mass per unit area from Smithwick et al. (2002) to compare
384 our measured wood loads in Washington to upland downed wood loads so as to examine both
385 how fluvial wood storage compares to upland downed wood storage and how logging affects that
386 comparison (Figure S2). We find that the two logged basins likely do not store more wood than
387 their corresponding uplands, whereas the unlogged wet fluvio-genic basin may store more wood
388 than nearby uplands.

388

389 In summary, logging has significantly decreased wood loads in the logged compared to
390 the unlogged wet fluvio-genic basin. Although logging has likely had a similar effect on the wet
391 glaciogenic basin, we cannot evaluate the local effects of logging on that basin.

391

3.2 Controls on the Proportion of Wood Stored in Jams

392

393 Despite wood jam density being a significant predictor in models of wood load, median
394 proportions of wood stored in jams for each basin are all well below 50% (Figure 2b, Table S2).
395 While some reaches store almost all wood as jams, wood is generally not stored as jams in these
396 dominantly small- to moderate-drainage area study reaches.

396

397 Multiple logistic regression modeling in the wet glaciogenic basin yields bankfull depth
398 and whether the reach is multithread (a measure of spatial heterogeneity) as significant predictors
399 of the proportion of wood in jams. Multithread channels are significantly more likely than single
400 thread channels to store wood as jams (wood is 1.05 to 23.16 times more likely to be stored in a
jam if the reach is multithread) and deeper channels tend to store more wood as jams than

401 shallower channels (wood is 0.94 to 6.12 times more likely to be stored in a jam for every 1 m
402 increase in bankfull depth).

403 In summary, we find that bankfull depth and channel planform control the proportion of
404 wood stored in jams in the wet glaciogenic basin.

405 **4 Discussion**

406 4.1 Inter-basin Comparisons and the Impacts of Logging on Wood Load

407 We compare wood loads between basins to examine the effects of climate (comparing the
408 semi-arid to the wet basins) and logging (comparing the logged and unlogged wet fluvio-genic
409 basins) on basin-scale wood load. Differences in wood loads between basins (Figure 2a) can be
410 largely explained by differences in precipitation and land use that result in differing forest stand
411 characteristics. The semi-arid glaciogenic basin, with the lowest wood loads, has the
412 correspondingly lowest precipitation and canopy cover ($p < 0.0001$ for comparisons with all
413 other basins). Mean canopy cover in the semi-arid glaciogenic basin is $27\% \pm 4\%$, whereas mean
414 canopy cover in the wet glaciogenic, unlogged wet fluvio-genic, and logged wet fluvio-genic
415 basins are $65\% \pm 5\%$, $73\% \pm 6\%$, and $72\% \pm 6\%$, respectively (uncertainty from a 95%
416 confidence interval on the mean). This likely indicates, and field observations support, that
417 forests in the semi-arid basin are less dense, trees are smaller, and the resulting supply of wood
418 to the channel is lower.

419 Although the mean slope of the basin upstream of each reach is not a mechanistic
420 predictor of hillslope instability, the semi-arid glaciogenic basin also has, on average, the lowest
421 upstream basin slopes compared to the wet basins ($p < 0.0001$ for comparisons with all other
422 basins). This indicates that landslides that could deliver large pulses of logs to channels are likely
423 much less frequent in the semi-arid basin compared to the wet basins. This is consistent with
424 estimates of upstream basin mean slope in the semi-arid basin being $17^\circ \pm 2^\circ$, whereas upstream
425 basin slopes in other basins generally hover around 30° ($29^\circ \pm 1^\circ$ in the wet glaciogenic, $31^\circ \pm 2^\circ$
426 in the logged wet fluvio-genic, and $29^\circ \pm 1^\circ$ in the unlogged wet fluvio-genic basin). Assuming
427 that a hillslope angle of around 30° is a threshold at which landslides become significantly more
428 frequent (Clarke & Burbank, 2010; Larsen & Montgomery, 2012), this indicates that basins in
429 the Pacific Northwest are likely experiencing relatively frequent landslides that potentially input
430 large pulses of logs to valley bottoms (Benda, Veldhuisen, et al., 2003; Benda & Bigelow, 2014).
431 In addition to the significantly denser forests and larger logs, the likelihood of more pulsed
432 inputs to channels in the Pacific Northwest probably explains higher wood loads. Although the
433 semi-arid basin has a lower wood jam density than all other basins ($p = 0.03$ compared to logged
434 wet fluvio-genic, 0.005 for unlogged wet fluvio-genic, and <0.0001 for wet glaciogenic), it is
435 unclear whether jams are simply less likely to form or whether lack of jams is a result of lower
436 wood loads, which is driven more by the lower supply of riparian trees to the channel.

437 Logging, in addition to climate, acts as an inter-basin scale control on wood load.
438 Comparing the three wet study basins, the unlogged fluvio-genic basin exhibits a significantly
439 higher wood load than the logged glaciogenic and fluvio-genic basins. The wet glaciogenic basin
440 exhibits much wider valley bottoms and larger drainage area than the fluvio-genic basins,
441 potentially confounding comparison. However, even when we restrict this comparison to reaches
442 with drainage areas lower than the maximum drainage area sampled in the unlogged wet
443 fluvio-genic basin (eliminating reaches with high drainage area and wide valley bottoms from the
444 wet glaciogenic basin), the unlogged wet fluvio-genic basin still exhibits significantly higher
445 wood loads than the wet glaciogenic basin ($p < 0.0001$) and likely higher wood loads than the

446 logged wet fluvio-genic basin ($p = 0.06$). This indicates that logging (as opposed to valley
447 morphology) is the dominant cause of reduced wood loads in the wet glaciogenic and logged wet
448 fluvio-genic basins, both of which exhibit statistically similar wood loads. Considering the similar
449 precipitation and forest stand characteristics throughout most of the basins (the exception being
450 the subalpine and alpine zones of the wet glaciogenic basin), and the observation that both are
451 extensively logged, it seems that wood loads in the wet glaciogenic and logged wet fluvio-genic
452 basins are likely lower as a direct result of logging.

453 We can use comparisons between the three basins in the Pacific Northwest to identify
454 likely mechanisms by which logging has reduced wood loads. While logging can enhance wood
455 supply to valley bottoms by increasing the frequency of landslides that deliver wood (Guthrie,
456 2002; Jakob, 2000; Roberts et al., 2004; Sidle et al., 2006; Wolter et al., 2010), it generally
457 reduces wood supply decreasing the quantity and size of trees available to be recruited to the
458 stream, especially in the absence of riparian buffers (Bilby & Ward, 1991; Ralph et al., 1994).
459 Logging also reduces in-channel and floodplain roughness if wood is removed or if streams are
460 cleared for tie-drives (anthropogenic floods that serve to flush wood down a channel after
461 harvest). This reduction in macro-scale roughness may reduce wood loads by reducing the
462 frequency of upstream-facing obstacles (e.g., bars, islands, large boulders) on which wood can be
463 trapped during high flows (Hyatt & Naiman, 2001; Ruffing et al., 2015; Wohl, 2014).

464 Our data indicates that a reduction in wood supply is the most likely mechanism by
465 which logging has reduced wood loads in the logged wet fluvio-genic basin, as opposed to a
466 reduction in tree size or a reduction in roughness due to tie-drives. Comparing wood sizes in the
467 logged and unlogged wet fluvio-genic basins, there are no significant differences in median ($p =$
468 0.43) or maximum ($p = 0.70$) piece diameter, or median ($p = 0.26$) or maximum ($p = 0.19$) piece
469 length, all of which could potentially relate to wood trapping efficiency. However, the logged
470 wet fluvio-genic basin consistently has a lower (albeit insignificantly different) estimated median
471 piece size and the possibility remains that wood pieces in the logged basin may be smaller than
472 the unlogged wet fluvio-genic basin. We observed no abandoned splash dams and there are no
473 recorded instances of tie-drives in the logged wet fluvio-genic basin, so logging probably did not
474 directly affect in-channel roughness. Jam density and the proportion of wood stored in jams does
475 not significantly differ between basins, suggesting that logging has not had a direct impact on the
476 storage patterns of wood in these rivers. We suspect that the combined effect of clearcut
477 harvesting reducing hillslope wood loads and harvest in the riparian zone reducing the supply of
478 wood to the channel results in lower wood loads. Our results indicate a similar reduction in wood
479 load by logging to what has previously been observed in northern wet conifer forests (Wohl,
480 Lininger, et al., 2017). Notably, our study examines the entirety of two otherwise nearly identical
481 basins, lending increased rigor to our comparison relative to past studies.

482 4.2 Controls on Wood Load

483 Our methodology in each basin differed, making generalization of these results difficult.
484 However, we can draw general conclusions across all basins by considering likely explanations
485 for observed intra-basin variability in wood load.

486 Jam density clearly controls wood load across all basins, despite the proportion of wood
487 stored in jams being significantly less than half in all basins. This indicates that despite their
488 relatively small proportion of storage, wood jams play a disproportionately large role in
489 determining total wood storage within a reach. This may be due to both the structure of wood
490 jams and their impacts on reach-scale wood mobility. Relatively stable wood jams are
491 hypothesized to significantly decrease the mobility of wood pieces in transport (Beckman &

492 Wohl, 2014; Kramer & Wohl, 2016). When analyzing the univariate relationship between jam
493 density and wood load in dispersed pieces, only data from the semi-arid glaciogenic basin
494 display a positive Spearman correlation ($p = 0.01$, $\rho = 0.34$ with a 95% confidence interval
495 between 0.06 and 0.57), weakly suggesting that pieces may be more likely to accumulate on jams
496 when more jams are present. We observe a significant positive correlation between jam density
497 and the proportion of wood stored in jams in all basins combined ($p < 0.0001$, $\rho = 0.95$ with a
498 95% confidence interval between 0.93 and 0.96), as well as in each individual region (all p
499 values < 0.0001 , 95% confidence intervals of ρ ranging from 0.74 to 1). This indicates that wood
500 pieces may preferentially deposit on existing accumulations as jam density increases.

501 The proportion of wood stored in jams in the wet glaciogenic basin is largely controlled
502 by planform and bankfull depth. We were unable to examine controls on the proportion of wood
503 stored in jams for other basins due to a lack of data (see section 2.5). It is likely that multithread
504 reaches, by having greater spatial heterogeneity in terms of flow depth variance and the presence
505 of bar heads and secondary channels, provide relatively immobile objects to anchor wood jams
506 and allow accumulation of racked pieces. This corroborates the interpretation of Wohl et al.
507 (2018), who found that the proportion of wood stored in jams is controlled mainly by whether
508 the reach contains multiple channels and Gurnell et al. (2000), who found that geomorphic
509 complexity directly related to wood retention within a reach. The effect of bankfull depth on the
510 proportion of wood stored in jams could be due to channels with greater bankfull depth being
511 able to transport larger logs at a given discharge, making individual pieces more mobile (Iroumé
512 et al., 2015; Kramer & Wohl, 2016). More mobile pieces transported past jams that are stable for
513 a given flow would likely lead to more wood stored in jams. Although wood jam stability
514 remains a major knowledge gap, our results indicate that spatial heterogeneity, specifically the
515 presence of upstream-facing surfaces on which wood can be trapped during high flows, appears
516 to regulate wood jam dynamics and, in turn, wood load.

517 The significance of elevation in determining wood load is likely due to trends in forest
518 type with elevation in the glaciogenic basins, as both basins have significant portions of the
519 stream network near and above tree line. As forests become thinner and trees grow more slowly
520 at higher elevations (see section 2.1), the supply of wood from hillslopes to the channel via mass
521 movement probably decreases, leading to a decrease in wood load. Conversely, the homogeneity
522 of forests in the fluvio-genic basins (likely due to the relatively low relief in those basins)
523 probably results in little variation in forest stand characteristics, explaining why elevation has no
524 significant effect on wood load in those basins.

525 In the fluvio-genic basins, we are surprised that slope, as opposed to bankfull channel
526 width or dimensionless piece length (L^*) significantly controlled wood load, since we tend to
527 observe what appear to be more dense accumulations of wood in smaller, steeper channels (e.g.,
528 Figure S3). Slope directly correlates to wood load in these basins and likely also directly
529 correlates to both channel width and the prevalence of large, relatively immobile roughness
530 elements (e.g., boulders) that can trap wood pieces. Higher gradient channels tend to have more
531 cascade or step-pool morphology and large boulders. These are largely absent from the lower
532 gradient portions of the network, which tend to erode either bedrock or gravel to cobble sized
533 substrate. Large clasts can interact with wood to form relatively stable accumulations in steeper
534 streams (Scott et al., 2014). This, combined with the fact that higher gradient reaches tend to
535 have narrower bankfull widths and corresponding valley widths ($p < 0.0001$, $\rho = -0.59$ with a
536 95% confidence interval between -0.75 and -0.36), probably leads to higher gradient reaches
537 being both able to trap wood in transport more effectively on large, relatively immobile

538 roughness elements and makes intact trees more likely to be able to span the channel, trapping
539 mobile wood until they begin to break down.

540 Confinement exerts a consistent and significant control on wood loads in both
541 glaciogenic basins. When wood pieces are able to interact with stable elements of hillslopes such
542 as living trees or stumps, they tend to resist mobilization (Beckman & Wohl, 2014b; Carah et al.,
543 2014, Figure S3). Such interaction is only possible if logs within the channel can reach such
544 elements on the hillside, which is more likely when channels are confined by their valley walls.
545 Unconfined reaches, especially those with less vegetated floodplains (observed in montane
546 meadows in the semi-arid glaciogenic basin or lower gradient reaches of the wet glaciogenic
547 basin with wide gravel bars) may be able to transport wood more readily without the wood being
548 trapped on floodplain or hillslope roughness elements.

549 It is notable that we are unable to find an effect of L^* on wood load, despite measuring
550 reaches spanning a range of L^* values from nearly 0 to 15. However, we find that the presence of
551 wood jams strongly controls wood loads, and the proportion of wood stored in jams is
552 dominantly a function of channel morphology, according to our modeling. Specifically, the
553 relationship between bankfull depth and the proportion of wood in jams may indicate that wood
554 mobility (regulated in part by bankfull depth) influences wood storage pattern. This indicates that
555 L^* alone may be insufficient to predict wood mobility. We find that the factors controlling wood
556 load at the reach scale do not appear to be as scale-dependent with respect to piece length and
557 channel width as has been hypothesized (Kramer & Wohl, 2016), but instead are relatively
558 consistent across the ranges of piece length to channel width examined here.

559 4.3 Conceptual Model of Wood Load in Rivers

560 We summarize our results and generalize them along with results from previous studies
561 in the form of a conceptual model (Figure 3) to describe the dominant controls on valley bottom
562 wood load at multiple spatial scales. While this conceptual model stems directly from our results,
563 we note that it represents a hypothesis that is explicitly tested by our analyses. We pose this
564 conceptual model to address the lack of a holistic conceptualization of the controls on wood
565 loads that applies to spatial scales from that of a single reach to entire watersheds or regions.
566 While previous work has suggested that quantifying wood load requires site-specific variables,
567 we instead argue that the following conceptual model should allow for these site-specific
568 variables to be viewed in a way that generalizes the processes affecting wood loads, enabling
569 future evaluation of multivariate models that accurately describe wood load in a variety of
570 settings and at multiple scales.

571 4.3.1 Wood Supply

572 Wood supply refers to the wood flux into the channel from mass movement (Benda &
573 Bigelow, 2014; Martin & Benda, 2001) and riparian recruitment via channel migration (Piégay et
574 al., 2017). The contribution of wood from mass movement depends on forest stand
575 characteristics (i.e., the amount of wood growing on hillsides) and the likelihood of mass
576 movements. Such mass movements are much more common in landscapes where hillslopes
577 reach a threshold mean gradient, proposed to be around 30° (Larsen & Montgomery, 2012), such
578 as those found in the Western Cordillera (Benda, Miller, et al., 2003; Benda & Bigelow, 2014).
579 However, mass movement likely contributes only a small proportion of wood flux to channels.
580 Wood likely comes more dominantly from riparian mortality (related to forest stand
581 characteristics and hydroclimatic/disturbance regimes) and bank erosion (Benda & Bigelow,
582 2014; Piégay et al., 2017). Our results indicating relationships between proxies for forest stand

583 density (elevation at an intra-basin scale and climate or logging at an inter-basin scale) and wood
584 load support the idea that land use and hydroclimatic regime determine forest characteristics and
585 resulting wood supply (Hough-Snee et al., 2015).

586 While our analysis does not directly examine recruitment rate, rates of lateral mobility
587 depend primarily on hydrology, geomorphology, and wood and vegetation dynamics (Brooks et
588 al., 2003; Collins et al., 2012; Richard et al., 2005; Wickert et al., 2013). Broadly, higher degrees
589 of spatial heterogeneity (i.e., multi-thread planforms, active lateral migration) may lead to higher
590 rates of wood supply to channels. At the same time, some forms of spatial heterogeneity
591 (discussed below) and recruitment can be direct results of in-channel and floodplain wood. In
592 this way, spatial heterogeneity, mainly channel morphology dynamics, links a feedback between
593 wood load and wood supply to channels (Figure 3).

594 4.3.2 Trapping Efficiency, a Combination of Storage Pattern and Spatial Heterogeneity

595 Our results indicate that jam density is a dominant control on wood load. In our
596 conceptual model, storage pattern refers to how wood is stored in the valley bottom: either on
597 floodplains or in the channel and either as jams or dispersed pieces. In addition, the breakdown
598 of wood by physical breakage or decay also influences how wood is stored, because these
599 processes regulate wood size (Gurnell, 2013). Storage pattern likely plays a strong role in
600 determining the stability of a piece of wood, or how long it will reside within a reach. Wood
601 stored on the floodplain should be more stable than wood stored in the channel, because
602 mobilization of floodplain wood requires a higher magnitude (and correspondingly less frequent)
603 flow (Wohl, Cadol, et al., 2018). Wood stored in a jam should be, on average, more stable than
604 dispersed pieces (Wohl & Goode, 2008), due to interactions among pieces of wood, sediment,
605 and in-channel and floodplain roughness elements (Bocchiola et al., 2008). Wood load directly
606 feeds back on storage pattern (Figure 3), as it is likely that a threshold wood load in channels is
607 required for the formation of jams. More work is needed to understand the mechanism by which
608 jam density relates to wood loads.

609 Spatial heterogeneity refers to floodplain and channel morphologic complexity and
610 ability to impede wood in transport. Essentially, a smooth, simplified channel with little
611 morphologic variability is less likely to provide features that can retain wood in transport than a
612 morphologically complex channel that exhibits upstream-facing surfaces on which wood can be
613 pinned. Such morphologic complexity can come from a variety of mechanisms. For instance,
614 large, relatively immobile boulders (Braudrick & Grant, 2000), living vegetation both within
615 channels (Dunkerley, 2014; Opperman et al., 2008) and on bars and floodplains, and vegetated
616 islands (Bertoldi et al., 2013; Gurnell et al., 2002) can all act as trapping points for wood in
617 transport. These objects can rack key pieces that can generate wood jams and can act as anchors
618 for dispersed pieces that impact them during transport. Heterogeneity in planform (e.g., bars and
619 pools, meanders) can result in wood deposition in shallower zones of flow in larger channels
620 (Gurnell et al., 2000; Wohl, Scott, et al., 2018). Channel geometry relative to wood length
621 (Kramer & Wohl, 2016; Shields et al., 2006) can determine how likely wood pieces are to span
622 the channel or ramp up on a bank (Wohl, 2013), increasing their resistance to mobilization.
623 While more spatially heterogeneous multithread channels do not significantly store more wood
624 in our modeling, we do find that multithread channels store higher proportions of wood in jams,
625 which may influence wood load via jam density.

626 Interpreting our results in the context of similar studies on larger rivers with wider
627 channels relative to log lengths reveals how stream size may influence the nature of spatial
628 heterogeneity. The small to medium streams studied here are generally more confined (i.e., logs

629 interact with banks frequently) and spatial heterogeneity is commonly in the form of bedform
630 variability, large boulders, and bankside vegetation that can trap wood ramped on floodplains
631 and valley walls. Larger streams display spatial heterogeneity dominantly in the form of bars and
632 mid-channel islands that generate shallow flow regions that tend to trap wood (Gurnell et al.,
633 2000; Wohl, Scott, et al., 2018). Our observed positive correlation between slope and wood load
634 in the fluvio-genic basins likely reflects the fact that streams in these basins are uniformly
635 confined by their valley walls, allowing bankside spatial disparities to trap wood, and making
636 large boulders or bedforms the dominant wood trapping mechanisms that can trap wood and
637 maintain jams (Scott et al., 2014). Such morphologic roughness features are likely more common
638 in higher gradient channels in those basins (Aberle & Smart, 2003). For the glaciogenic basins,
639 the relationship between slope and wood load is insignificant, likely reflecting the fact that both
640 boulders, bankside disparities, and bedforms as well as planform irregularity, bars, and in-stream
641 vegetation contribute to wood trapping. In those basins, more confined reaches likely allow
642 wood to interact more strongly with bankside heterogeneities, leading to high wood loads.

643 Vegetation patch dynamics regulate riparian forest stand characteristics (a feedback
644 between spatial heterogeneity and wood supply) as well as the potential for wood to be impeded
645 in transport, especially on bar or floodplain surfaces (Fetherston et al., 1995). Wood in the
646 channel can determine vegetation patch dynamics by affecting the formation of hard points in the
647 valley bottom (Collins et al., 2012), acting as a feedback between wood load and spatial
648 heterogeneity (Figure 3). Lateral mobility is a function of both how effective the river is at
649 eroding its banks and depositing bars as well as the limitations exerted by valley walls or
650 anthropogenic confinement. Our observation that confinement is a strong control on wood load,
651 whereby more confined channels have higher wood loads (Wyżga et al., 2017), however,
652 suggests that greater lateral mobility may result in decreased wood trapping efficiency, despite
653 potential increases in recruitment rate. The exception to this may be found in the case of larger
654 rivers (Gurnell et al., 2000; Wohl, Scott, et al., 2018), where wider reaches may have more bars
655 and islands on which wood can be retained.

656 With our conceptual model, we propose that wood load is a function of how much wood
657 is deposited within a reach and its residence time, and is controlled by characteristics that affect
658 storage patterns, spatial heterogeneity, and the supply of logs to the channel. Together, spatial
659 heterogeneity and storage pattern determine trapping efficiency, or the wood retentiveness of a
660 reach. This conceptual model relates these characteristics to wood load and facilitates discussion
661 of how wood load feeds back on storage pattern and spatial heterogeneity, which in turn feeds
662 back on supply.

663 4.4 Valley Bottom Wood Contribution to the Riverine OC Pool

664 A recent compilation of wood OC storage in temperate rivers shows that, with one
665 exception, most past quantifications of wood OC stock are in the range of 1 to 150 Mg C/ha
666 (Sutfin et al., 2016). Comparing the first-order estimates from our study basins to other values
667 from temperate regions contextualizes the impact of logging on the wood OC stock. In the semi-
668 arid glaciogenic basin, with much of its area near or above tree line, wood plays a minor role in
669 storing carbon (95% confidence interval on median between 0.0 and 2.5 Mg C/ha). In contrast,
670 wet basins in the Pacific Northwest demonstrate substantial OC storage in the form of wood
671 (95% confidence interval on median between 2.7 and 27.9 Mg C/ha). Notably, wood OC storage
672 in the unlogged wet fluvio-genic basin (95% confidence interval on median between 67.4 and
673 229.5 Mg C/ha) is high compared to most temperate rivers (Sutfin et al., 2016), many of which
674 have been impacted by anthropogenic wood removal or a loss of wood supply (Wohl, 2014;

675 Wohl, Lininger, et al., 2017). This highlights the potential wood OC storage contribution of
676 undisturbed temperate watersheds. The factor of 2 decrease in wood load between the unlogged
677 and logged wet fluvio-genic basins in the context of the large extent of anthropogenic
678 disturbances to mountain river basins implies that wood OC storage in mountain river basins has
679 been significantly impacted by anthropogenic disturbance, and that restoration of wood load may
680 have a significant impact on valley bottom OC storage (Lininger et al., 2017).

681 Understanding the spatial variability in wood residence times is now essential to guide
682 wood load management in the context of climate change and efforts to retain carbon on the
683 landscape. While most wood found in channels is likely less than 50 years old, wood stored in
684 floodplains can reach ages on the order of $10^2 - 10^3$ years (Guyette et al., 2002, 2008; Hyatt &
685 Naiman, 2001; Nanson et al., 1995; Webb & Erskine, 2003). Despite this high variability, wood
686 is likely a significant contribution to the valley bottom carbon pool (Naiman et al., 1987; Sutfin
687 et al., 2016; Wohl et al., 2012). It is important to better quantify how long the substantial riverine
688 wood OC pool resides on the landscape, and its eventual fate after it leaves a watershed (either
689 by export or decay). For example, in the case of the Olympic mountains, it is unknown whether
690 wood is more recalcitrant in mountain river basins or as driftwood in the near-shore environment
691 (Schwabe et al., 2015; Simenstad et al., 2003).

692 **5 Conclusions**

693 We present quantifications of wood load across the entirety of four river basins across the
694 western U.S. to understand intra- and inter-basin variability in wood load spatial distribution.
695 Our modeling shows that wood jam density, confinement, elevation, and slope are strong
696 controls on wood loads. Comparing basins with differing land use and those with differing
697 climate reveals the strong impact of wood supply on wood loads.

698 Interpreting these results in the context of past studies allows us to conceptualize wood
699 load through the interaction of wood supply to the valley bottom and the efficiency of the valley
700 bottom at trapping wood delivered to it (Figure 3). We find that differences in wood load
701 between basins with varying precipitation and forest stand characteristics are likely the result of
702 factors influencing wood supply. Local geomorphic factors such as wood storage pattern and
703 valley bottom morphology best explain reach-scale variation in wood load. This implies that
704 wood load modeling must take into account effects operating at varying spatial scales.
705 Importantly, our results suggest that after accounting for basin-scale variation in variables such
706 as precipitation and forest characteristics, relatively consistent factors control wood load at the
707 reach-scale, namely those that describe spatial heterogeneity and wood storage pattern. We
708 hypothesize that while every basin is different (Hough-Snee et al., 2015), future multivariate
709 predictive models based on this multi-scale conceptualization of wood load controls will likely
710 be able to accommodate inter-basin variability and predict wood load at the reach scale in a
711 variety of hydroclimatic regions. All factors influencing wood supply and trapping efficiency
712 listed in Figure 3 are quantifiable in both field and flume environments. As such, future
713 statistical analyses, predictive modeling, and experimentation should be able to use the
714 conceptual model we propose as a starting point for determining relevant variables across spatial
715 scales to be used in multivariate modeling of wood load.

716 The factor of two difference between wood loads in the logged and unlogged wet
717 fluvio-genic basins demonstrates the severe impact of clearcut logging with no riparian buffer and
718 provides a clear representation of the potential enhancement of the river corridor that could be
719 achieved by watershed-scale restoration. Restoration actions currently underway in the logged
720 wet fluvio-genic basin (Pacific District Olympic National Forest, 2012) focus on addressing the

721 wood supply deficiency that likely causes this wood-poor state. However, our conceptual model
722 suggests that addressing the wood supply impacts of logging at the basin scale will likely only be
723 successful if trapping efficiency is addressed, such that wood is retained within the basin. On a
724 positive note, our comparisons do not suggest that the valley bottom morphology or the density
725 of wood jams differs significantly between these two basins, indicating that the logged wet
726 fluviogenic basin may have similar trapping efficiency to the unlogged wet fluviogenic basin.

727 In terms of OC storage in valley bottoms, we demonstrate that, especially in wood-rich,
728 undisturbed river networks, wood provides a high magnitude pool of OC. This OC pool may
729 persist for 10^3 years (Guyette et al., 2002; Hyatt & Naiman, 2001), although wood residence time
730 is a major knowledge gap.

731 **Acknowledgements**

732 This work was funded by NSF grant EAR-1562713 and a National Geographic Society
733 Young Explorer Grant. We thank the Quileute tribe and Olympic National Park for access to
734 field sites in the Olympics. We thank Ellen Daugherty for extensive assistance in field work and
735 Katherine Lininger for stimulating discussion that improved the manuscript. Comments from
736 three anonymous reviewers and Charles Luce significantly improved the manuscript. All data
737 supporting the analyses presented here can be found in Dataset S1 as well as the CSU Digital
738 Repository (<https://hdl.handle.net/10217/186057>, Scott & Wohl, 2018).

739

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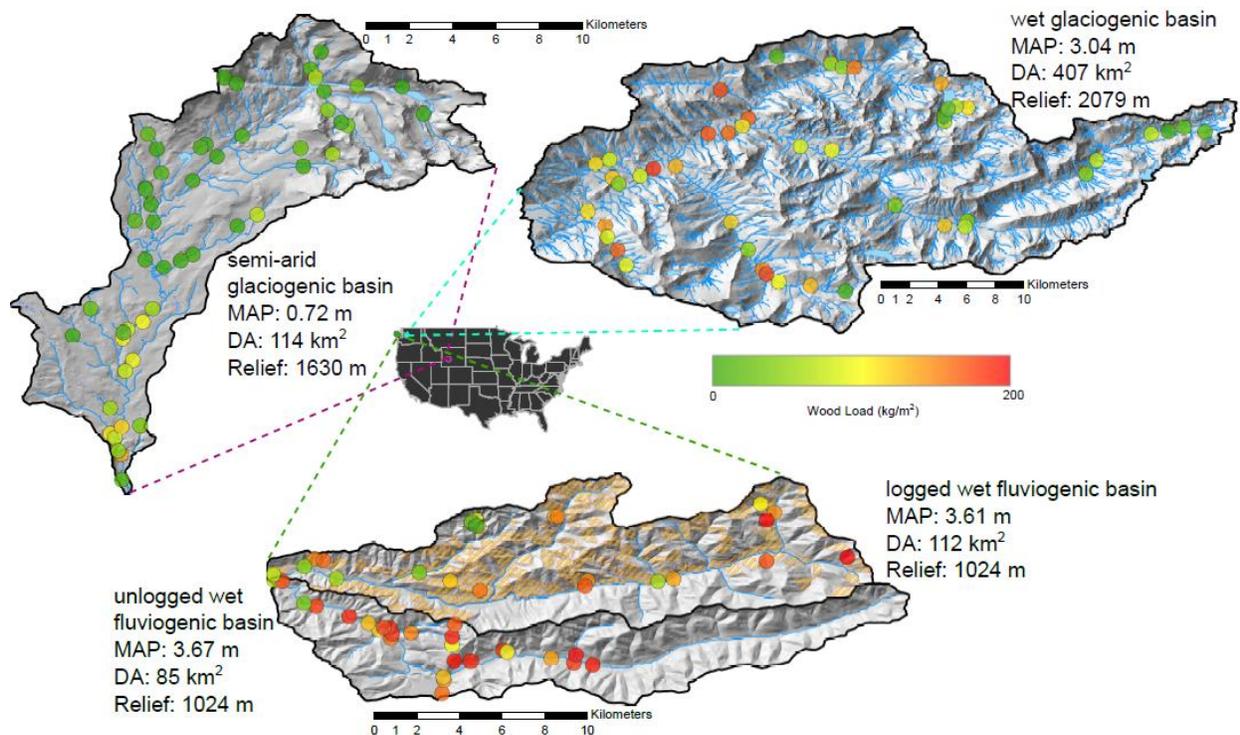
1034 **Tables**

1035 Table 1: Characteristics of study basins. Mean annual precipitation data are from PRISM
 1036 (Oregon State University, 2004). Relief, drainage area, and mean basin slope are calculated from
 1037 a 10 m DEM.

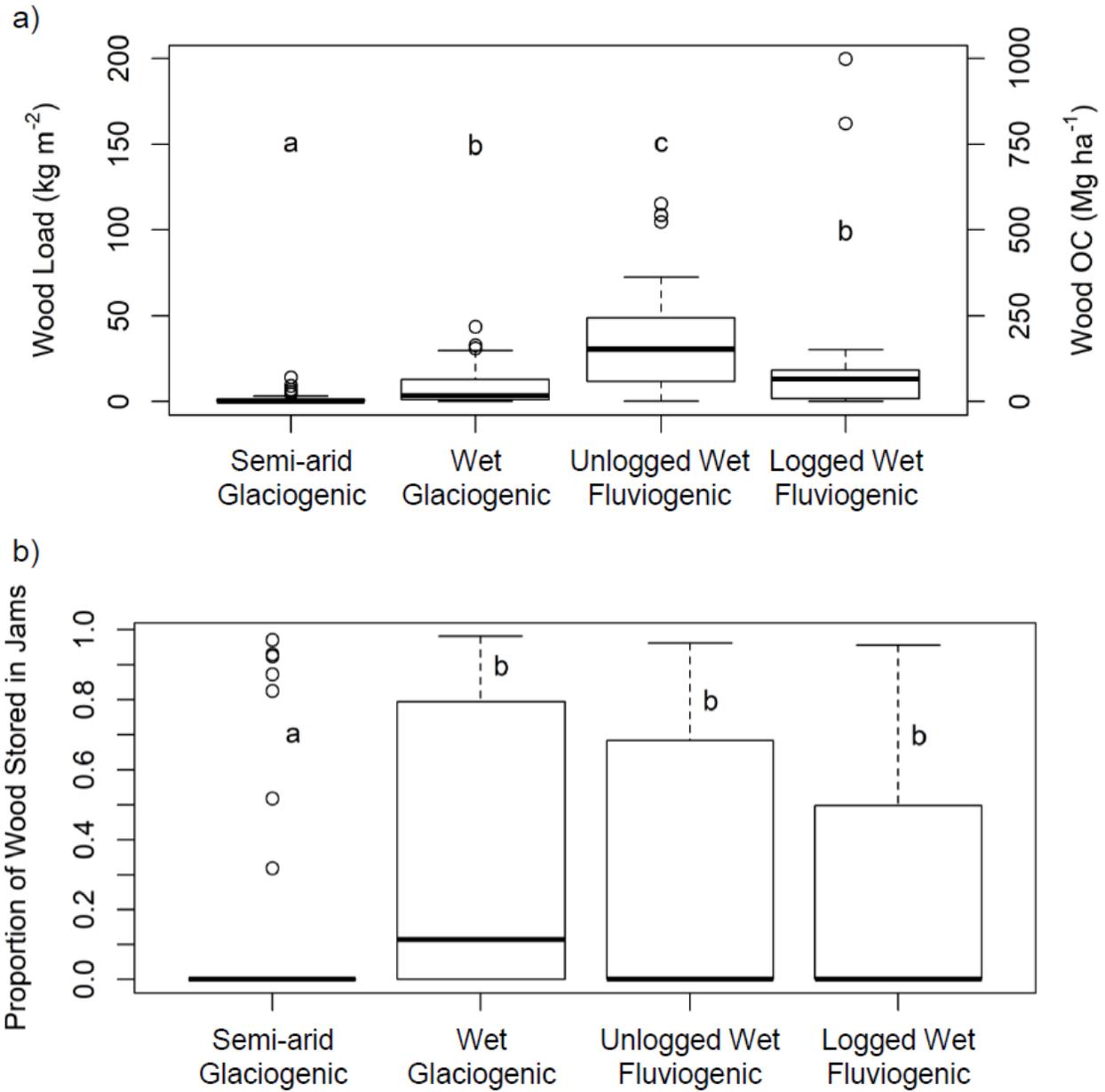
Basin	Mean Annual Precipitation (m)	Relief (m)	Drainage Area (km²)	Mean Basin Slope (%)
Logged Wet Fluviogenic	3.61	1024	112	49
Unlogged Wet Fluviogenic	3.67	1024	85	45
Wet Glaciogenic	3.04	2079	407	60
Semi-arid Glaciogenic	0.72	1630	114	25

1039 **Figures**

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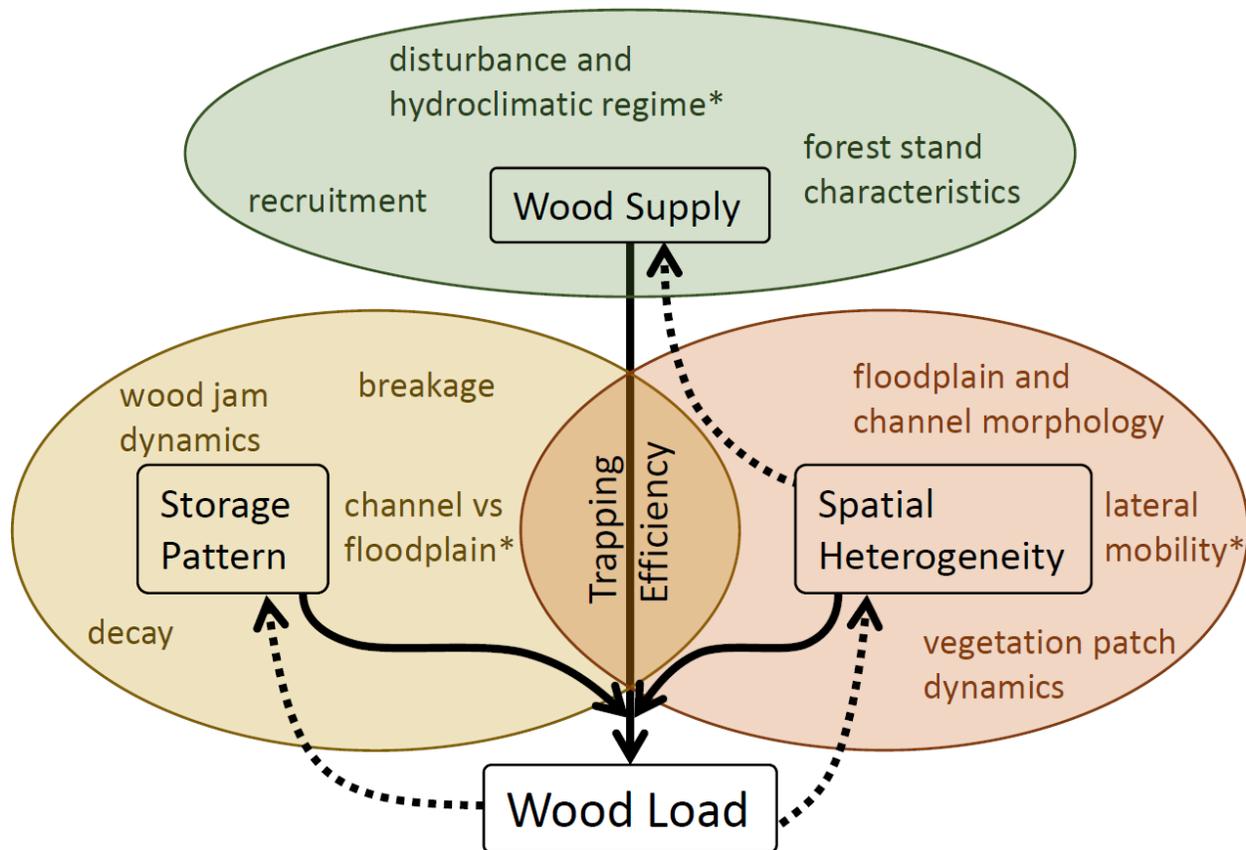


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 1042 Figure 1: Map showing the location, topography, sampling sites, and stream network of the
 1043 sampled basins. Clockwise, from upper left: Big Sandy watershed, Wyoming; MF Snoqualmie
 1044 watershed, Washington; Sitkum (north) and SF Calawah (south) watersheds, Washington.
 1045 Circles represent sampling locations at which wood loads were measured, and are colored by
 1046 wood load. The orange overlay in the Sitkum basin shows areas that have experienced recorded
 1047 clearcut timber harvest. Mean annual precipitation (MAP), drainage area (DA), and relief are
 1048 given for each basin.
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Figure 2: Boxplot of wood load and OC storage in wood (a) and the proportion of wood stored in jams (b) by study basin. Bold line represents median. Box top and bottom represent 75th and 25th percentile, respectively. Ends of dashed lines represent 1.5 times the interquartile range. Circles represent outliers. Letters show significantly different groups at a 95% confidence level. Data shown here are summarized in Table S2, and translated to wood volume for comparison in Figure S1.



1059
 1060 Figure 3: Conceptual model of controls on valley bottom wood load. Colored text within the
 1061 ellipse surrounding each control indicates the processes that regulate that control. Dotted arrows
 1062 represent feedbacks. Asterisks indicate processes that may determine other processes within each
 1063 ellipse. Wood supply regulates wood load through the filter of trapping efficiency. That is,
 1064 trapping efficiency is the first-order, local control on wood load, whereas wood supply is a
 1065 broader, basin-scale limit on maximum wood load. This model can be used to explain differences
 1066 in wood loads between basins (mainly related to wood supply), the effects of anthropogenic
 1067 activities or changing climate, and variation within a single basin. See section 4.3 for details.