

- Climate change causes river network contraction and disconnection in
 the H.J. Andrews Experimental Forest, Oregon, USA
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13 Abstract

- 14 Headwater streams account for more than 89% of global river networks and provide numerous
- 15 ecosystem services that benefit downstream ecosystems and human water uses. It has been
- 16 established that changes in climate have shifted the timing and magnitude of observed precipitation,
- 17 which, at specific gages, have been directly linked to long-term reductions in large river discharge.
- 18 However, climate impacts on ungaged headwater streams, where ecosystem function is tightly
- 19 coupled to flow permanence along the river corridor, remain unknown due to the lack of data sets and
- ability to model and predict flow permanence. We analyzed a network of 10 gages with 38 to 69
- 21 years of records across a 5th-order river basin in the U.S. Pacific Northwest, finding increasing
- 22 frequency of lower low-flow conditions across the basin. Next, we simulated river network
- expansion and contraction for a 65-yr period of record, revealing 24% and 9% declines in flowing
 and contiguous network length, respectively, during the driest months of the year. This study is the
- 24 and contiguous network length, respectively, during the driest months of the year. This study is the 25 first to mechanistically simulate network expansion and contraction at the scale of a large river basin,
- 26 informing if and how climate change is altering connectivity along river networks. While the
- heuristic model presented here yields basin-specific conclusions, this approach is generalizable and
- transferable to the study of other large river basins. Finally, we interpret our model results in the
- 29 context of regulations based on flow permanence, demonstrating the complications of static
- 30 regulatory definitions in the face of non-stationary climate.

31 1 Introduction

- 32 More than 89% of the global river network is headwaters (Allen et al., 2018; Downing et al., 2012),
- 33 supporting ecosystem services and the health of downstream waters (Alexander, Boyer, Smith,
- 34 Schwarz, & Moore, 2007; US EPA, 2015). These services are associated with the frequency with
- 35 which streams have surface flow (hereafter "flow permanence"), and any declines in flow
- 36 permanence will effectively disconnect larger rivers from their headwaters and their functions. Flow
- 37 generated in headwater streams is highly sensitive to changes in precipitation timing, magnitude, and

38 duration based on a small number of empirical studies over short timescales (Godsey & Kirchner,

2014; Jensen, McGuire, & Prince, 2017; Prancevic & Kirchner, 2019; Zimmer & McGlynn, 2017).

40 However, no observational studies have covered a sufficient period of record to evaluate if and how

- 41 changing climate has altered flow permanence across river networks. Consequently, numerical
- 42 simulations parameterized with readily available data are needed to fill this knowledge gap (Gallart et
- 43 al., 2016; Ward, Schmadel, & Wondzell, 2018a).

44 Changes in flow permanence can alter the transport and transformation of water, energy, dissolved

45 and suspended materials, and organisms throughout the river network (Datry, Bonda, & Boulton,

46 2017; Datry, Pella, Leigh, Bonada, & Hugueny, 2016; Gallart et al., 2012; Larned, Datry, Arscott, &

Tockner, 2010; Raymond, Saiers, & Sobczak, 2016; Steward, Von Schiller, Tockner, Marshall, &
Bunn, 2012). Evaluating how flow permanence has changed requires quantification of both the

Bunn, 2012). Evaluating how flow permanence has changed requires quantification of both the
temporal variation (i.e., the frequency a given segment has surface flow) and spatial variation (i.e.,

50 the spatial connectivity of surface flow) (Covino, 2017; Wohl, 2017; Wohl, Magilligan, & Rathburn,

51 2017). In headwater streams, flow permanence is controlled by the dynamic interaction of geologic

52 setting with hydrologic forcing (Costigan, Jaeger, Goss, Fritz, & Goebel, 2016; Prancevic &

53 Kirchner, 2019). Climate change is primarily associated with changes to hydrologic forcing, such as

altering the spatial distribution and within-year timing of precipitation. Geologic setting – such as

valley width and slope, sinuosity, and hydraulic conductivity – will remain relatively static compared

56 to the pace of climate change.

57 Changes in flow permanence complicate management and protection of headwater streams.

58 Regulatory protections in the U.S. and E.U. are traditionally focused on perennially flowing waters,

59 with emerging attention paid to temporarily flowing waters (Fritz et al., 2018; Nikolaidis et al., 2013;

60 US DoD, 1986; US DoD & EPA, 2015; US DoD & US EPA, 2018; Walsh & Ward, 2019). Further

61 complicating management, data on headwater streams, and particularly intermittent and ephemeral

62 streams, are lacking. For example, only 3% of the rivers gaged in the U.S. are headwater streams, as

gages are heavily biased toward larger rivers (Eng & Milly, 2007; Poff, Bledsoe, & Cuhaciyan,
 2006). A proposed rule would revise protected status to waters with contiguous surface flow in a

65 "typical" 30-year period in the U.S., but does not address systematic changes in flow permanence

66 (US DoD & US EPA, 2018). The time-variable definition of the 30-year window does not consider

67 the role of climate change and shifting norms, despite clear evidence that non-stationarity is prevalent

68 in hydrologic systems (Milly et al., 2008). For example, systematic declines in streamflow, and

69 particularly lower low-flows, have been observed across the U.S. Pacific Northwest (Luce and

70 Holden, 2009). Subsequent study reveals that slowing westerlies during winter months have reduced

the orographic enhancement of precipitation in the region, changing both the amount of precipitation

reaching the landscape and the timing for storage vs. export from catchments (Luce, Abatzoglou, &

Holden, 2013). A more recent example paints another dire picture for the future of streamflow in the

74 Southwestern U.S. in response to shifting precipitation and temperature (Milly & Dunne, 2020).

75 Here we assess whether flow permanence in headwater streams has shifted over the past 65 years

76 from the mid 20th century baseline in response to observed changes in climate-driven hydrologic

77 forcing. We investigate how timing and magnitude of discharge have shifted over a 65-y period of

record and yielded changes in flow permanence along mountain stream networks. Finally, we

consider how our findings may inform current and future protections for streams under the proposed

80 Waters of the U.S. (WOTUS) Rule (US DoD & US EPA, 2018) and subsequently finalized at the

81 Navigable Waters Protection Rule (US DoD & US EPA, 2019). We selected the 5th-order Lookout

82 Creek basin (Western Cascade Mountains, Oregon, USA) because of the extensive and long-term

83 network of gages on low-order streams (Table S2). Furthermore, this basin is representative of the

- 84 broader Pacific Northwest where climate change impacts on the timing and magnitude of moisture
- delivery to high elevation watersheds are known to cause declines in large rivers (Luce et al., 2013: 85
- 86 Luce & Holden, 2009). Thus, reduced orographic enhancement of precipitation due to climate change is projected at the field site. This study considers the cascading impact of this change on stream
- 87
- discharge, and how discharge changes in headwaters may change flow permanence and connectivity 88 in a river network. 89

90 2 Methods

91 2.1 Site description & available data

- The study was conducted at the H.J. Andrews Experimental Forest (HJA), a 5th-order basin in the 92
- Western Cascades, Oregon, USA (site map in Fig. S1). The basin drains about 6,400 ha of forested 93
- 94 landscape, with elevations ranging from about 410 to 1,630 m a.m.s.l, making it an ideal place to 95
- evaluate the impact of a changing climate on river networks of the broader Pacific Northwest. The 96 basin has been a long-term study site for ecological and forest management research for more than 70
- 97 years and is relatively pristine with no urban land use, no dams or reservoirs, and minimal logging
- 98 during the period of this study. At the longest currently operating meteorological station (CS2MET,
- 99 elev. 485 m a.m.s.l.) annual precipitation averages 2,345 mm and average annual air temperature
- 100 averages 9.2 deg. C. Additional summaries of temperature and precipitation including trends for the
- period of record for each station are summarized in Table S1 and Figures S2-S15). In general, 101
- 102 significant trends in monthly precipitation and air temperature are infrequently detected, due largely
- 103 to the short observational records at the local meteorological network (Table S1). Further details
- 104 about the local climate, morphology, geology, and ecology are comprehensively described elsewhere
- 105 (Cashman, Deligne, Gannett, Grant, & Jefferson, 2009; Deligne et al., 2017; Dyrness, 1969;
- 106 Jefferson, Grant, & Lewis, 2004; Swanson & James, 1975; Swanson & Jones, 2001).
- 107
- The HJA includes a network of 10 stream gages with drainage areas ranging from 8.5 to 6,241.9 ha, 108
- 109 with records of 38 to 69 years of data across the gage network (Table S2). Additionally, high quality
- 110 digital elevation model derived from an airborne LiDAR survey is available for the entire basin,
- 111 which has been reliably processed to extract topographic metrics including valley width, valley slope,
- 112 and stream sinuosity (after Corson-Rikert, Wondzell, Haggerty, & Santelmann, 2016; Schmadel,
- 113 Ward, & Wondzell, 2017b; Ward et al., 2018a; Ward, Schmadel, & Wondzell, 2018b). For each gage
- 114 we also calculated annual summary metrics of discharge including annual minimum discharge, mean
- 115 discharge, maximum discharge, exceedance discharges (1, 5, 10, 25, 50, 75, 90, 95, 99th percentiles),
- total annual discharge, and the days elapsed to various cumulative fractions of discharge (1, 5, 10, 25, 116
- 117 50, 75, 90, 95, 99%).
- 118
- 119 Finally, an extensive data collection effort spanning the stream orders and lithologic regions of the
- 120 basin was completed in 2015, providing a database of stream and valley morphologies and hydraulic
- 121 conductivities to inform network-scale model parameterization (Ward, Zarnetske, et al., 2019).

122 2.2 Simulation of the river network

- 123 Simulation of network expansion and contraction followed the methods, data sources, and
- 124 implementation described by *Ward et al.* (2018a). Briefly, the approach conceptualizes the river

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125 corridor in 1-D along the valley, with domains representing the surface stream and down-valley flow

126 in the valley bottom. Critically, the down-valley subsurface flow (or "underflow") is filled first, with

127 surface flow representing only the volumetric flow in excess of what the valley bottom can transmit

- downstream (Fig. 1). Put another way, surface flow occurs only when the valley subsurface cannot
- accommodate the down-valley discharge.

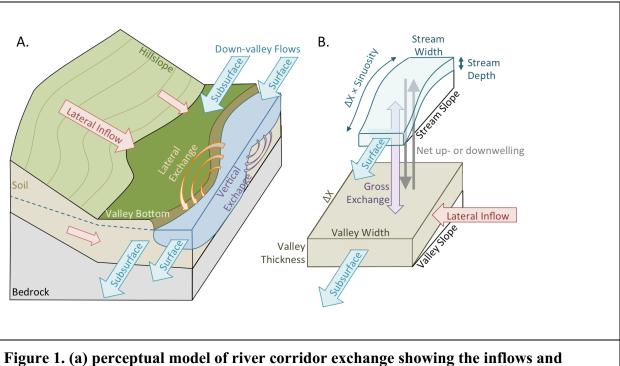


Figure 1. (a) perceptual model of river corridor exchange showing the inflows and outflows considered in the model. (b) representation of stores, fluxes, and key variables used in the model implementation. The key advance of this model is the representation of the river corridor as parallel transport in the subsurface and surface domains, with surface flow only activating when the subsurface cannot accommodate all down-valley flow. Reprinted with permission from Ward et al. (2018a).

- 130 *Ward et al.* (2018a) validated the model in a 2nd order catchment of our study basin, concluding the
- 131 model was appropriate to represent network expansion and contraction based on correct prediction of
- flowing or dry streambed conditions for more than 95% of over 3.2 million observations. We proceed
- 133 with implementation of this model across a 5th order basin on the basis of *Ward et al.* 's (2018a)
- success within our study basin, particularly given the accuracy in representing network expansion
- 135 and contraction in response to diurnal fluctuations driven by evapotranspiration, storms, and seasonal
- 136 baseflow.
- 137 The model was intentionally derived and constructed to require geomorphic and hydrologic data that
- are readily estimated for unstudied catchments (Ward et al., 2018a), consistent with our application in
- this study. Valley width, valley slope, along-stream slope, sinuosity, and the lateral contributing area
- 140 for each 10-m segment of the valley bottom were extracted from the LiDAR data using a modified
- 141 version of the TopoToolbox (Schmadel et al., 2017b; Schwanghart & Kuhn, 2010; Schwanghart &
- 142 Scherler, 2014). Stream width at each location was estimated using a power-law regression of 62
- 143 observations in August 2015 (Ward, Zarnetske, et al., 2019) as:

144

$b = 0.003926 * UAA^{0.4488}$

where b is the width of the channel (m) and UAA is the total drainage area (ha), and the best-fit 145

146 relationship had a coefficient of determination of 0.84. We assigned a uniform Manning's *n* of 0.045

along the entire network based on visual inspection during our experience working in the basin. Ward 147 148 et al. (2018a) identified hydraulic conductivity as the largest source of uncertainty in the model. In

149 response, we observed hydraulic conductivity of the streambed at 57 sites in August 2015 (Ward,

150 Zarnetske, et al., 2019) and assigned the geometric mean, 1.53×10^{-4} m s⁻¹, across the network.

151 Porosity was assigned as 0.3 at all locations, the midpoint of past studies (Dyrness, 1969; Kasahara &

152 Wondzell, 2003; Ward, Schmadel, Wondzell, Gooseff, & Singha, 2017; Wondzell, LaNier, &

153 Haggerty, 2009) and the same value used in model validation (Ward et al., 2018a). We set valley

154 colluvium depth to a minimum value of 1-m (Gooseff, Anderson, Wondzell, LaNier, & Haggerty,

155 2006; Ward et al., 2018a), increasing as:

h = 1 + 0.01w

157 where h is colluvium depth (m) and w is valley width (m). This functional form was selected to

158 reflect the limited measurements of subsurface colluvium depth that are available, including

159 geophysical observations at several headwater locations (Crook et al., 2008; Ward et al., 2012) and

along the 5th order reach of Lookout Creek (*Wondzell*, personal communication and unpublished 160

161 data).

156

162 To parameterize the total down-valley discharge at the upstream end of each 1st order segment, we

163 calculated a unique power-law regression for the gage discharge and drainage area for each 15-

164 minute timestep simulated, and defined the discharge based on UAA. Thus, all available gage data,

165 and their time variation, informed the upstream boundaries for model headwaters. Lateral inflows for

166 each segment were estimated using the same power-law regression, where the change in UAA

between the up- and downstream end of each segment was used to calculate the associated change in 167

discharge attributed to the lateral area (Schmadel et al., 2017b; Ward et al., 2018a, 2018b). Finally, 168

169 we used the threshold of at least 2.21×10^{-4} m³ s⁻¹ to differentiate surface flow from dry streambeds

170 after past studies in the basin (Ward et al., 2018a).

171 Consistent with Ward et al. (2018a), we underscore that reduced complexity models are intended to

172 represent the dominant mechanisms and interactions in a system of interest. This necessarily comes

173 at the expense of representation of complexity and heterogeneity within the system. While our model

174 has been derived and validated for headwaters within the study basin, the parameterization detailed

175 above requires simplifications. To that end, this model is most appropriately viewed as heuristic,

176 consistent with common practice in the study of river corridors (e.g., Cardenas & Wilson, 2007;

177 Gooseff et al., 2006; Irvine & Lautz, 2015; Schmadel, Ward, Lowry, & Malzone, 2016; Schmadel,

178 Ward, & Wondzell, 2017a; Trauth, Schmidt, Maier, Vieweg, & Fleckenstein, 2013). At the scale of

179 river networks, comparable models have been applied to study patterns and trends at large spatial

180 scales at the expense of site-specific localized predictions (Gomez-Velez & Harvey, 2014; e.g.,

Gomez-Velez, Harvey, Cardenas, & Kiel, 2015; Kiel & Cardenas, 2014; Schmadel et al., 2018). 181

182 Thus, given the model's strong performance at the reach-scale within our study basin (Ward et al.,

183 2018a), explicit design as a heuristic that can be implemented at minimally studied sites, the wealth of data available across our network, and the tradition of heuristic models to test hypotheses in river

184

185 corridor science, we proceed with this approach.

186 2.3 **Statistical tests**

- 187 All between-group differences were tested using one-way ANOVA, Kruskal-Wallis, and Mann-
- 188 Whitney-Wilcoxson U tests. We report differences as significant only if p < 0.05 for all three tests.
- 189 For all trends in discharge metrics, flowing frequency, contiguous frequency, flowing length, and
- 190 contiguous length, we used Mann-Kendall tests and Sen's slopes to define the significance and
- 191 direction of trends. Decreasing trends are reported for p < 0.05 for the Mann-Kendall tests and a
- 192 Sen's slope of less than zero. Increasing trends are reported for p < 0.05 for the Mann-Kendall tests
- and a Sen's slope greater than zero. We report no significant trend for a Mann-Kendall test with p > 100
- 194 0.05 or if Sen's slope is zero.
- 195 Analysis of trends may be sensitive to the length of the data set and which years or timesteps are
- 196 included (i.e., different starting or ending dates, or different trend lengths). Consequently, we
- analyzed trends for every moving window of 10 or more years for every metric considered in the
- 198 study, including those related to discharge at gages as well as flowing and contiguous lengths. In the 199 body of manuscript we report significance and direction based on overall trends for each analysis. In
- body of manuscript we report significance and direction based on overall trends for each analysis. In the supplemental information we also tabulate how many of the moving windows agree with the
- 201 overall trends, the length of the single longest trend that opposes the overall trend, and the length of
- the single longest period with no significant trend. We also tabulated the mean, median, maximum,
- and minimum Sen's Slope for every analysis, and the number of trends that are increasing,
- decreasing, or exhibit no significant trend. (see Table S4 for robustness of discharge trends, Table S5
- for robustness of flowing length and connected length trends, and Figs. S29-S30 for visualization of
- 206 the annual flowing length and connected length trends).
- 207 3 Results & Discussion

208 **3.1** Headwater stream discharge is declining during the dry season

- 209 Discharge is predominantly decreasing across all gages in the basin over a 65-yr period of record
- 210 (Fig. 2). For example, the Lookout Creek gage at the basin outlet has decreasing discharge for 81%
- of the year (about 300 days), steady discharge for 4% of the year (about 15 days), and increasing
- discharge for 15% of the year (about 55 days) (Fig. 2C; remaining sites in Figs S16-24; Tables S2-3).
- 213 The largest and most consistently decreasing trends are during the summer season when discharge is
- 214 lowest. We found no increasing discharge for the driest 7-months of the year (April through October;
- 215 Fig. 2D).
- 216 Across the gage network, we find significant inter- and intra-annual changes in the timing and
- 217 magnitude of discharge. Annual mean, median, and total discharge are all decreasing for 9 of 10
- 218 gages across their periods of record. We also found decreasing annual minimum and maximum
- discharges for 7 of 10 stream gages, declining annual low-flows (75-99% exceedance flows) at all
- 220 gages, and declining annual high-flows (1- 25% exceedance flows) at 7 of 10 gages (Table S3).
- 221 Conceptually, the changes in moisture delivery is causing an increased export of water during winter
- 222 months (Luce et al., 2013; Table S3), as evidenced by the more rapid time to export the first 10% of
- streamflow each year. Consequently, less water is stored during the rainy season, resulting in
- decreased dry-season baseflow and extended times to export the last 10% of annual discharge.

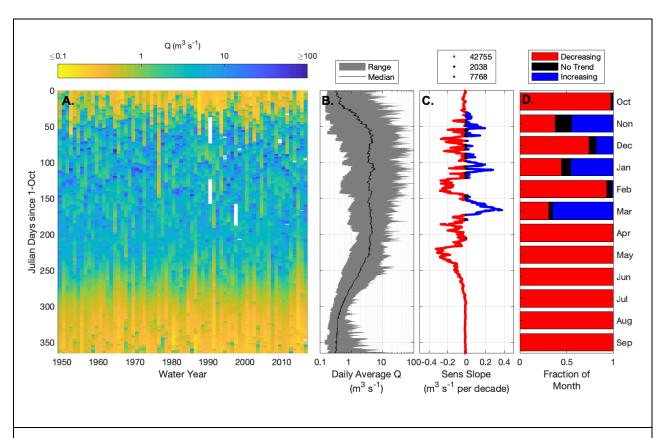


Figure 2. 10-min discharge analysis for Lookout Creek gage near the outlet of the basin. (A) discharge for every day, every year of record (white indicates missing data). (B) Range of discharges from the period of record. The median discharge for each 10-min timestep and the hydrograph from the wettest and driest years (i.e., those with the largest and smallest total annual discharge) are displayed atop the range. (C) Sen's slope for each 10-min period. Color indicates trend is significantly decreasing (red; 81% of the year; n = 42,755; p < 0.05), not changing (black; 4% of the year; n = 2,038), or significantly increasing (blue; 15% of the year; n = 7,768; p < 0.05). (D) Fraction of each month with decreasing, no, or increasing discharge trends for the period of record. See Supplemental Figs. S6-S24 for comparable plots of all gages in this study.

225

226 **3.2** Decreased flow permanence has reduced river network connectivity

227 Using the stream gage data, topographic analysis, and published data collected in the basin, we 228 simulated dynamic expansion and contraction of the network (Fig.3, Fig. S25) (Ward et al., 2018a). 229 For the 65-yr simulation period, declining discharge and increasing early season export of water 230 within the basin result in an overall contraction of the flowing network (Fig. 3). We found the 231 flowing network reaches a maximum length of about 40 km during the wet winter months and contracts to as short as 15-km during the driest periods of record (Fig. 3B, Fig. S25). Flowing 232 233 network length is a useful proxy for connectivity along the river corridor, where longer flowing 234 lengths allow more rapid connection of hillslopes to downstream water and promote rapid export of 235 energy and materials rather than internal transformation (van Meerveld, Kirchner, Vis, Assendelft, & 236 Seibert, 2019).

- 237 Next, we found the connected network length plateaus at about 21-km during the wet winter months
- and contracts to as small as about 5-km under the lowest discharge conditions (Fig. S10). The
- connected network represents an average of 57.4% of the flowing network across the entire
- simulation (median 55.9%; range 8.9% to 79.8%). The connected network defines the migration
- corridor through which aquatic organisms may travel upstream from the basin outlet without
- encountering a dry streambed location.

243 We found significant declines in flowing length for 75.7% of the year (about 276 days) compared to

244 23.6% of the year with no-trend (about 86 days), and less than 1% of the year (about 3 days) with

- increasing flowing length (Fig. 3C). The decreasing trends are common throughout much of the year
- except for the highest discharge conditions associated with spring storms and snowmelt runoff (April
- through June) when the network length is more steady. Connected length exhibits similar trends,
 declining for 66.7% of the year (about 243 days), no trend for 33.1% of the year (about 121 days) and
- increasing trends for less than 1% of the year (about 2 days), no trend for 33.1% of the year (ab 249 increasing trends for less than 1% of the year (about 2 days; Fig. S25).
- 250 Decreasing flowing and contiguous lengths are not distributed evenly through the year. Flowing
- length declines by a long-term average rate of 21.0 m yr⁻¹ (median 2.3 m yr⁻¹ contraction; range 124.3 m yr⁻¹ contraction to 1.1 m yr⁻¹ expansion) and connected length declines by an average of 4.7 m yr^{-1}
- 252 (median 0.6 m yr⁻¹ decline; range 44.1 m yr⁻¹ decline to 0.40 m yr⁻¹ expansion). The largest average
- rate of flowing length decline, 94.2 m yr⁻¹, is in September. Average September flowing length is
- 255 24.1% shorter in in 2009-2018 than 1953-1962. Similarly, connected length averages a loss of 21.4 m
- 256 yr⁻¹ in August for 2009-2018, and is 9.2% shorter than the 1953-1962 August average.
- 257 Network expansion and contraction exhibit threshold behavior, generally consistent with past studies
- 258 (Prancevic & Kirchner, 2019; Ward et al., 2018a). When discharge at the Lookout Creek gage is
- 259 greater than about 1 m³ s⁻¹, the flowing and connected lengths are nearly constant at their plateau (T_{1}, T_{2})
- values (Fig. 3B, Fig. S27). Under these wet-season, high discharge conditions, the flowing length
- 261 maximum reflects a geologic limitation on network expansion where the drainage network is
- sufficiently dense to drain additional precipitation from the landscape without developing additional channels. As discharge drops below $1 \text{ m}^3 \text{ s}^{-1}$, the network dynamically expands and contracts in
- response to precipitation. Under these dynamic conditions, the capacity of the valley bottom is
- comparable to the down-valley discharge, resulting in the large variation in flowing length in
- response to minor fluctuations in discharge(Ward et al., 2018a).

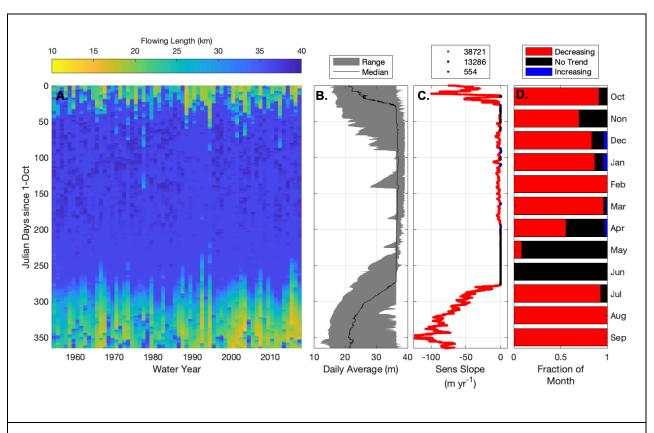


Figure 3. Summary of flowing lengths for the simulation period, including: (A) flowing length for every 10-minute period for water years 1953 through 2017; (B) annual flowing lengths; (C) Sen's slope for the period of record, calculated for every 10-min period of the year; and (D) fraction of each month that length is decreasing, increasing, or has no significant trend. A comparable plot of contiguously flowing length is provided in Fig. S25.

267

3.3 Headwater streams in steeper and/or wider valleys are the most sensitive to climate change

About 41% of the headwater stream network exhibit a decreasing surface flow frequency, with the

remaining 59% exhibiting no change (Fig. 4A). No location had increasing frequency of surface

flow. Similarly, 27% of locations decrease in frequency of connected flow, 73% have no change, and

273 no sites are more frequently connected across the period of record (Fig. S11).

274 Declining trends in flowing and connected frequency are not evenly distributed through the year.

275 Instead, we found few significant trends for any segment during the wet season (November through

June) because maximum network extent is controlled by basin morphology and drainage density

277 (Fig. 4B). During the dry season (July – October) we found declining frequency of surface flow and

contiguous flow in many network segments, due to declining discharge during this period. Similarly,

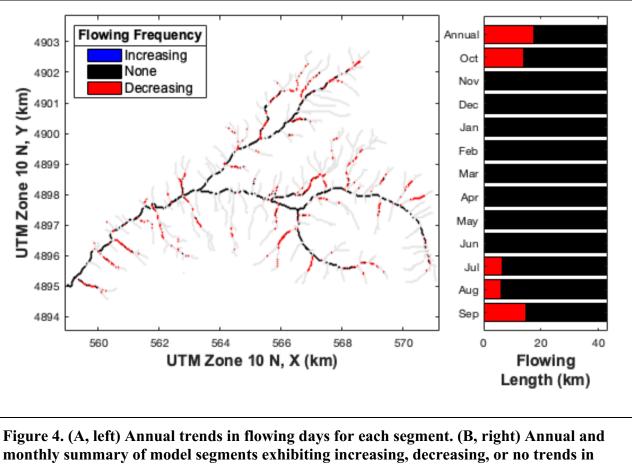
trends in flowing and connected frequency are not evenly distributed in space. The reaches with the

280 largest declines in flowing and connected frequency have significantly smaller drainage areas, steeper

- valley slopes, and/or wider valleys compared to locations with no trend (Fig. S28), consistent with
- past findings in smaller catchments and the conceptual model (Costigan, Daniels, & Dodds, 2015;
- 283 Prancevic & Kirchner, 2019; Ward et al., 2018a). Decreasing trends in both flowing and connected

284 frequency are most prevalent at the most upstream extents of the network, making the lowest-order

streams "canaries in the coal mine" to first detect the impacts of climate change on flow permanence.



monthly summary of model segments exhibiting increasing, decreasing, or no trends in flowing frequency. Panel B presents data discretized in space, while Fig. 3D presents data lumped in time. A comparable plot for trends in contiguously flowing status is presented in Fig. S26.

286

287 **3.4** Changing flow permanence challenges current regulatory strategies

Non-stationarity is now the dominant paradigm in water resources (Milly et al., 2008). In our study system, the peak and average connected lengths are significantly larger in the first 30 years than the last 30 years. From a practical perspective, some waters that would have been federally jurisdictional (here for finite dictional?) in 1082 (here does the period 1052, 1082) means the invited integral in 2018

- 291 (herafter "jurisdictional") in 1982 (based on the period 1953-1982) may not be jurisdictional in 2018
- 292 (based on the period 1989-2018).
- 293 The proposed definition for Waters of the United States, which defines the basis for a water receiving
- 294 federal protections under the federal Clean Water Act, focuses on the frequency of surface flow (US
- 295 DoD & US EPA, 2018). The definition would establish jurisdiction over streams with surface water
- flow in a "typical year" based on precipitation during a rolling 30-yr window absent extreme flood
- and drought events. Thus, flow permanence is a de facto standard for protections, but its systematic
- 298 changes with climate are not accounted for in regulations. For example, if jurisdictional status is

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defined as flow during 29 of 30 years, jurisdictional network length would decline from about 29-km

- in 1982 to about 26-km in 2017 (horizontal range at Y = 29-yr; Fig. 5). As the minimum number of
- flowing years for regulatory protection decreases, changes due to climate become negligible (e.g.,
- horizontal range at Y = 15-yr; Fig. 5). In contrast, if the Navigable Waters Protection Rule was intended to provide a constant-in-time determination, it must explicitly adjust the definition of
- 304 "typical year" with climate. The systematic contraction in our study system, and thereby loss of
- federal protections for streams and their nearby wetlands, is only one response to changes in climate.
- 306 In a landscape where flow permanence increases due to changing climate, federal jurisdictional scope
- 307 could increase. Our critique here is consistent with draft comments from the US EPA's Science
- 308 Advisory Board (Honeycutt & Board, 2019).

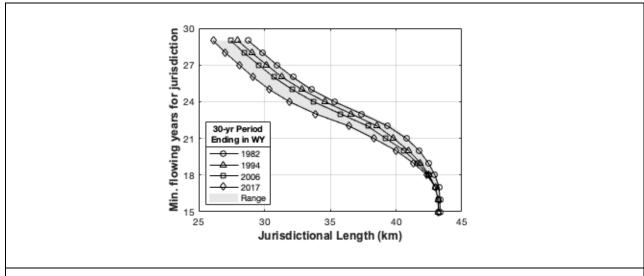


Figure 5. Federal jurisdictional length (x-axis) as a function of the minimum number of flowing years to establish jurisdiction (y-axis) and final year of rolling 30-yr period. The horizontal range between the lines is the change in jurisdictional length for a fixed definition of "typical year" (e.g., must flow 25 of 30 years to be jurisdictional). The vertical range between lines is the reduction in threshold that would be required to protect the same length of stream.

309

310 4 Conclusions

311 While past studies have explored the reduction in discharge at downstream gages on large rivers

- 312 (Luce & Holden, 2009), this study is the first to examine how known changes in precipitation (Luce
- et al., 2013) and discharge translate into changes in connectivity between mountain hillslopes and
- their headwaters. Compared to their 1953-1962 averages, the 2009-2018 network has contracted by
- 315 24.1% and 9.2% in flowing and connected length, respectively, during the driest months. The
- 316 dynamic connections along the network underpin a host of ecosystem services that we expect to also
- vary with flowing frequency. The loss of ecological function of such streams could be irreversible,
- 318 and time-variable jurisdictional protections complicate the protection of these important resources.
 319 These losses are relative to a mid 20th century baseline, and while some function will be lost as flow
- These losses are relative to a mid 20th century baseline, and while some function will be lost as flow permeance decreases, other functions could be amplified as a result of increased duration or
- 321 frequency of non-flowing conditions. Decreases in streamflow during periods when water resources

- are in highest demand recently observed across the western U.S. (e.g., Milly & Dunne, 2020) further
- 323 pinpoints the need for extending the approaches presented here to more river basins.
- 324 Simulations predict that reaches with smaller drainage areas and larger subsurface flow capacity are
- 325 the most likely to change in their flowing and connected frequencies in response to climate change.
- 326 Thus headwater locations with steep valleys gradients, larger valley widths, and/or disproportionately
- high hydraulic conductivity (Ward, Wondzell, et al., 2019; Wondzell, 2006) should be closely
- 328 monitored to assess catchment response to climate change. Importantly, there are a small number of 329 critical locations within a valley that can cutoff entire upstream reaches from the contiguous network
- 329 critical locations within a valley that can cutoff entire upstream reaches from the contiguous network
 330 one location with a wider- or steeper-than-average morphology can transition to entirely subsurface
- 331 flow. We observed this threshold disconnection when the Lookout Creek gage discharge below about
- $1 \text{ m}^3 \text{ s}^{-1}$. While this threshold is the result of local geologic setting, but we expect other systems will
- exhibit similar threshold behavior as a function subsurface flow capacity and discharge. Finally, we
- underscore that current regulations are not designed with climate change-induced shifts in flowing
- and connected frequency, which will complicate policy enforcement for protection of headwater
- 336 streams. The conclusions presented here are specific to one river basin in the Pacific Northwest, but
- the modeling approach and interpretation were intentionally designed to be transferable to other river
- networks, enabling extended analysis with modest, commonly-available data.

339 5 Conflict of Interest

The authors declare that the research was conducted in the absence of any commercial or financialrelationships that could be construed as a potential conflict of interest.

342 6 Author Contributions

ASW conceived of the study and led the modeling, analysis, and writing. All authors participated in

344 the data analysis and writing of the study. NMS completed the topographic analyses and contributed

to conceptual and numerical model development. ASW and SMW secured funding to support the

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- 358 Any use of trade, firm, or product names is for descriptive purposes only and does not imply
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360 9 References

- Alexander, R. B., Boyer, E. W., Smith, R. A., Schwarz, G. E., & Moore, R. B. (2007). The Role of
 Headwater Streams in Downstream Water Quality. *JAWRA Journal of the American Water Resources Association*, 43(1), 41–59.
- Allen, G. H., Pavelsky, T. M., Barefoot, E. A., Lamb, M. P., Butman, D., Tashie, A., & Gleason, C.
 J. (2018). Similarity of stream width distributions across headwater systems. *Nature Communications*, 9(1), 610. https://doi.org/10.1038/s41467-018-02991-w
- Cardenas, M. B., & Wilson, J. L. (2007). Exchange across a sediment-water interface with ambient
 groundwater discharge. *Journal of Hydrology*, *346*(3–4), 69–80.
- Cashman, K. V., Deligne, N. I., Gannett, M. W., Grant, G. E., & Jefferson, A. (2009). Fire and water:
 Volcanology, geomorphology, and hydrogeology of the Cascade Range, central Oregon. *Field Guide*, 15, 539–582. https://doi.org/10.1130/2009.fl
- Corson-Rikert, H. A., Wondzell, S. M., Haggerty, R., & Santelmann, M. V. (2016). Carbon dynamics
 in the hyporheic zone of a headwater mountain streamin the Cascade Mountains, Oregon. *Water Resources Research*, *52*, 7556–7576. https://doi.org/10.1029/2008WR006912.M
- Costigan, K. H., Daniels, M. D., & Dodds, W. K. (2015). Fundamental spatial and temporal
 disconnections in the hydrology of an intermittent prairie headwater network. *Journal of Hydrology*, 522, 305–316. https://doi.org/10.1016/j.jhydrol.2014.12.031
- Costigan, K. H., Jaeger, K. L., Goss, C. W., Fritz, K. M., & Goebel, P. C. (2016). Understanding
 controls on flow permanence in intermittent rivers to aid ecological research: integrating
 meteorology, geology and land cover. *Ecohydrology*, 9(7), 1141–1153.
 https://doi.org/10.1002/eco.1712
- Covino, T. (2017). Hydrologic connectivity as a framework for understanding biogeochemical flux
 through watersheds and along fluvial networks. *Geomorphology*, 277, 133–144.
 https://doi.org/10.1016/j.geomorph.2016.09.030
- Crook, N., Binley, A. M., Knight, R., Robinson, D. a., Zarnetske, J. P., & Haggerty, R. (2008).
 Electrical resistivity imaging of the architecture of substream sediments. *Water Resources Research*, 44(4), n/a-n/a. https://doi.org/10.1029/2008WR006968
- Datry, T., Bonda, N., & Boulton, A. J. (Eds.). (2017). *Intermittent Rivers and Ephemeral Streams : Ecology and Management* (1st ed.). London, U.K.: Academic Press.
- Datry, T., Pella, H., Leigh, C., Bonada, N., & Hugueny, B. (2016). A landscape approach to advance
 intermittent river ecology. *Freshwater Biology*, *61*(8), 1200–1213.
 https://doi.org/10.1111/fwb.12645
- Deligne, N. I., Mckay, D., Conrey, R. M., Grant, G. E., Johnson, E. R., O'Connor, J., & Sweeney, K.
 (2017). Field-trip guide to mafic volcanism of the Cascade Range in Central Oregon—A
 volcanic, tectonic, hydrologic, and geomorphic journey. *Scientific Investigations Report*, 110.
 https://doi.org/10.3133/sir20175022H
- Downing, J. A., Cole, J. J., Duarte, C. M., Middelburg, J. J., Melack, J. M., Prairie, Y. T., ... Tranvik,
 L. J. (2012). Global abundance and size distribution of streams and rivers. *Inland Waters*, 2(4),

- 399 229–236. https://doi.org/10.5268/IW-2.4.502
- 400 Dyrness, C. T. (1969). Hydrologic properties of soils on three small watersheds in the western
 401 Cascades of Oregon. USDA FOREST SERV RES NOTE PNW-111, SEP 1969. 17 P.
- 402 Eng, K., & Milly, P. C. D. (2007). Relating low-flow characteristics to the base flow recession time
 403 constant at partial record stream gauges. *Water Resources Research*, 43(1), 1–8.
 404 https://doi.org/10.1029/2006WR005293

Fritz, K. M., Schofield, K. A., Alexander, L. C., McManus, M. G., Golden, H. E., Lane, C. R., ...
Pollard, A. I. (2018). Physical and Chemical Connectivity of Streams and Riparian Wetlands to
Downstream Waters: A Synthesis. *Journal of the American Water Resources Association*, 54(2),
323–345. https://doi.org/10.1111/1752-1688.12632

Gallart, F., Llorens, P., Latron, J., Cid, N., Rieradevall, M., & Prat, N. (2016). Validating alternative
methodologies to estimate the regime of temporary rivers when flow data are unavailable. *Science of the Total Environment*, 565, 1001–1010.

412 https://doi.org/10.1016/j.scitotenv.2016.05.116

Gallart, F., Prat, N., Garca-Roger, E. M., Latron, J., Rieradevall, M., Llorens, P., ... Froebrich, J.
(2012). A novel approach to analysing the regimes of temporary streams in relation to their
controls on the composition and structure of aquatic biota. *Hydrology and Earth System Sciences*, *16*(9), 3165–3182. https://doi.org/10.5194/hess-16-3165-2012

Godsey, S. E., & Kirchner, J. W. (2014). Dynamic, discontinuous stream networks: Hydrologically
 driven variations in active drainage density, flowing channels and stream order. *Hydrological Processes*, 28(23), 5791–5803. https://doi.org/10.1002/hyp.10310

Gomez-Velez, J. D., & Harvey, J. W. (2014). A hydrogeomorphic river network model predicts
where and why hyporheic exchange is important in large basins. *Geophysical Research Letters*,
422 41, 6403–6412. https://doi.org/doi:10.1002/2014GL061099

- Gomez-Velez, J. D., Harvey, J. W., Cardenas, M. B., & Kiel, B. (2015). Denitrification in the
 Mississippi River network controlled by flow through river bedforms. *Nature Geoscience*,
 8(October), 1–8. https://doi.org/10.1038/ngeo2567
- Gooseff, M. N., Anderson, J. K., Wondzell, S. M., LaNier, J., & Haggerty, R. (2006). A modelling
 study of hyporheic exchange pattern and the sequence, size, and spacing of stream bedforms in
 mountain stream networks, Oregon, USA. *HYDROLOGICAL PROCESSES*, 20(11), 2443–2457.
- Honeycutt, M., & Board, U. E. S. A. (2019). Draft Commentary on the Proposed Rule Defining the
 Scope of Waters Federally Regulated Under the Clean Water Act.
- 431 Irvine, D. J., & Lautz, L. K. (2015). High resolution mapping of hyporheic fluxes using streambed
 432 temperatures: Recommendations and limitations. *Journal of Hydrology*, *524*, 137–146.
 433 https://doi.org/10.1016/j.jhydrol.2015.02.030
- Jefferson, A., Grant, G. E., & Lewis, S. L. (2004). A River Runs Underneath It: Geological Control
 of Spring and Channel Systems and Management Implications, Cascade Range, Oregon.
- 436 Advancing the Fundamental Sciences Proceedings of the Forest Service: Proceedings of the

- 437 *Forest Service National Earth Sciences Conference*, *1*(October 2004), 18–22.
- Jensen, C. K., McGuire, K. J., & Prince, P. S. (2017). Headwater stream length dynamics across four
 physiographic provinces of the Appalachian Highlands. *Hydrological Processes*, *31*(19), 3350–
 3363. https://doi.org/10.1002/hyp.11259
- Kasahara, T., & Wondzell, S. M. (2003). Geomorphic controls on hyporheic exchange flow in
 mountain streams. *Water Resources Research*, *39*(1), 1005.
- Kiel, B., & Cardenas, M. (2014). Lateral hyporheic exchange throughout the Mississippi River
 network. *Nature Geoscience*, 7(May), 413–417. https://doi.org/10.1038/ngeo2157
- Larned, S. T., Datry, T., Arscott, D. B., & Tockner, K. (2010). Emerging concepts in temporary-river
 ecology. *Freshwater Biology*, 55(4), 717–738. https://doi.org/10.1111/j.13652427.2009.02322.x
- Luce, C. H., Abatzoglou, J., & Holden, Z. A. (2013). The Missing Mountain Water : Slower
 Westerlies Decrease Orographic Enhancement in the Pacific Northwest USA. *Science*, *1360*(2013), 1360–1365. https://doi.org/10.1126/science.1242335
- Luce, C. H., & Holden, Z. A. (2009). Declining annual streamflow distributions in the Pacific
 Northwest United States, 1948-2006. *Geophysical Research Letters*, *36*(16), 2–7.
 https://doi.org/10.1029/2009GL039407
- Milly, P. C. D., Bentacourt, J., Falkenmark, M., Robert, M., Hirsch, R. M., Kundzewicz, Z. W., ...
 Stouffer, R. J. (2008). Stationarity is dead: Whither water management? . *Science*, *319*(5863),
 573–574.
- Milly, P. C. D., & Dunne, K. A. (2020). Colorado River flow dwindles as warming-driven loss of
 reflective snow energizes evaporation. *Science*, *9187*(February), eaay9187.
 https://doi.org/10.1126/science.aay9187
- Nikolaidis, N. P., Demetropoulou, L., Froebrich, J., Jacobs, C., Gallart, F., Prat, N., ... Perrinl, J. L.
 (2013). Towards sustainable management of Mediterranean river basins: Policy
 recommendations on management aspects of temporary streams. *Water Policy*, *15*(5), 830–849.
 https://doi.org/10.2166/wp.2013.158
- Poff, L. N., Bledsoe, B. P., & Cuhaciyan, C. O. (2006). Hydrologic variation with land use across the
 contiguous United States: geomorphic and ecological consequences for stream ecosystems.
 Geomorphology, 79(3), 264–285.
- 467 Prancevic, J. P., & Kirchner, J. W. (2019). Topographic Controls on the Extension and Retraction of
 468 Flowing Streams. *Geophysical Research Letters*, 46(4), 2084–2092.
 469 https://doi.org/10.1029/2018GL081799
- 470 Raymond, P. A., Saiers, J. E., & Sobczak, W. V. (2016). Hydrological and biogeochemical controls
 471 on watershed dissolved organic matter transport: pulse- shunt concept. *Ecology*, 97(1), 5–16.
- Schmadel, N. M., Harvey, J. W., Alexander, R. B., Schwarz, G. E., Moore, R. B., Eng, K., ... Scott,
 D. (2018). Thresholds of lake and reservoir connectivity in river networks control nitrogen

- 474 removal. *Nature Communications*, *9*(1). https://doi.org/10.1038/s41467-018-05156-x
- Schmadel, N. M., Ward, A. S., Lowry, C. S., & Malzone, J. M. (2016). Hyporheic exchange
 controlled by dynamic hydrologic boundary conditions. *Geophysical Research Letters*, 43,
 4408–4417. https://doi.org/10.1002/2016GL068286.Received
- 478 Schmadel, N. M., Ward, A. S., & Wondzell, S. M. (2017a). Hydrologic controls on hyporheic
 479 exchange in a headwater mountain stream. *Water Resources Research*.
 480 https://doi.org/10.1002/2017WR020576
- 481 Schmadel, N. M., Ward, A. S., & Wondzell, S. M. (2017b). Hydrologic controls on hyporheic
 482 exchange in headwater mountain streams. *Water Resources Research*, 53(7), 6260–6278.
 483 https://doi.org/10.1002/2017WR020576
- 484 Schwanghart, W., & Kuhn, N. J. (2010). TopoToolbox: A set of Matlab functions for topographic
 485 analysis. *Environmental Modelling and Software*, 25(6), 770–781.
 486 https://doi.org/10.1016/j.envsoft.2009.12.002
- 487 Schwanghart, W., & Scherler, D. (2014). Short Communication: TopoToolbox 2 MATLAB-based
 488 software for topographic analysis and modeling in Earth surface sciences. *Earth Surface*489 *Dynamics*, 2(1), 1–7. https://doi.org/10.5194/esurf-2-1-2014
- 490 Steward, A. L., Von Schiller, D., Tockner, K., Marshall, J. C., & Bunn, S. E. (2012). When the river
 491 runs dry: Human and ecological values of dry riverbeds. *Frontiers in Ecology and the*492 *Environment*, 10(4), 202–209. https://doi.org/10.1890/110136
- 493 Swanson, F. J., & James, M. E. (1975). *Geology and geomorphology of the H.J. Andrews* 494 *Experimental Forest, western Cascades, Oregon.* Portland, OR.
- Swanson, F. J., & Jones, J. a. (2001). *Geomorphology and Hydrology of the H.J. Andrews Experimental Forest, Blue River, Oregon.* 1–5.
- 497 Trauth, N., Schmidt, C., Maier, U., Vieweg, M., & Fleckenstein, J. H. (2013). Coupled 3-D stream
 498 flow and hyporheic flow model under varying stream and ambient groundwater flow conditions
 499 in a pool-riffle system. *Water Resources Research*, 49(9), 5834–5850.
 500 https://doi.org/10.1002/wrcr.20442
- 501 US DoD. (1986). Final Rule for Regulatory Programs of the Corps of Engineers. *Federal Register*,
 502 51(219), 41206–41260. https://doi.org/10.1002/0471686786.ebd0198.pub2
- 503 US DoD, & EPA, U. (2015). Clean Water Rule: Definition of "Waters of the United States." *Federal* 504 *Register*, 80(124), 37054–37127.
- 505 US DoD, & US EPA. (2018). *Revised Definition of "Waters of the United States"* (pp. 253 pp. EPA-506
 HQ-OW-2018–0149). pp. 253 pp. EPA-HQ-OW-2018–0149.
- 507 US DoD, & US EPA. (2019). Definition of "Waters of the United States" Recodification of Pre 508 Existing Rules (Vol. 84).
- 509 US EPA. (2015). Connectivity of Streams & Wetlands to Downstream Waters: A Review and

- 510 Synthesis of the Scientific Evidence (Final Report).
- 511 https://doi.org/10.1017/CBO9781107415324.004
- van Meerveld, H. J. I., Kirchner, J. W., Vis, M. J. P., Assendelft, R. S., & Seibert, J. (2019).
 Expansion and contraction of the flowing stream network changes hillslope flowpath lengths
 and the shape of the travel time distribution. *Hydrology and Earth System Sciences Discussions*,
 2006(September 2006), 1–18. https://doi.org/10.5194/hess-2019-218
- Walsh, R., & Ward, A. S. (2019). Redefining Clean Water Regulations Reduces Protections for
 Wetlands and Jurisdictional Uncertainty. *Frontiers in Water*, 1(April), 1–6.
 https://doi.org/10.3389/frwa.2019.00001
- Ward, A. S., Fitzgerald, M., Gooseff, M. N., Voltz, T. J., Binley, A. M., & Singha, K. (2012).
 Hydrologic and geomorphic controls on hyporheic exchange during base flow recession in a headwater mountain stream. *Water Resources Research*, 48(4), W04513.
- Ward, A. S., Schmadel, N. M., & Wondzell, S. M. (2018a). Simulation of dynamic expansion,
 contraction, and connectivity in a mountain stream network. *Advances in Water Resources*, *114*,
 64–82. https://doi.org/10.1016/j.advwatres.2018.01.018
- Ward, A. S., Schmadel, N. M., & Wondzell, S. M. (2018b). Time-Variable Transit Time
 Distributions in the Hyporheic Zone of a Headwater Mountain Stream. *Water Resources Research*. https://doi.org/10.1002/2017WR021502
- Ward, A. S., Schmadel, N. M., Wondzell, S. M., Gooseff, M. N., & Singha, K. (2017). Dynamic
 hyporheic and riparian flow path geometry through base flow recession in two headwater
 mountain streamcorridors. *Water Resources Research*, *53*.
 https://doi.org/10.1002/2016WR019875
- Ward, A. S., Wondzell, S. M., Schmadel, N. M., Herzog, S., Zarnetske, J. P., Baranov, V., ...
 Wisnoski, N. I. (2019). Spatial and temporal variation in river corridor exchange across a 5th
 order mountain stream network. *Hydrology and Earth System Sciences Discussions*, (April), 1–
 https://doi.org/10.5194/hess-2019-108
- Ward, A. S., Zarnetske, J. P., Baranov, V., Blaen, P. J., Brekenfeld, N., Chu, R., ... Wondzell, S. M.
 (2019). Co-located contemporaneous mapping of morphological, hydrological, chemical, and
 biological conditions in a 5th order mountain stream network, Oregon, USA. *Earth System Science Data Discussions*, (April), 1–27. https://doi.org/10.5194/essd-2019-45
- 540 Wohl, E. (2017). Connectivity in rivers. *Progress in Physical Geography*, *41*(3), 345–362.
 541 https://doi.org/10.1177/0309133317714972
- Wohl, E., Magilligan, F. J., & Rathburn, S. L. (2017). Introduction to the special issue: Connectivity
 in Geomorphology. *Geomorphology*, 277, 1–5. https://doi.org/10.1016/j.geomorph.2016.11.005
- Wondzell, S. M. (2006). Effect of morphology and discharge on hyporheic exchange flows in two
 small streams in the Cascade Mountains of Oregon, USA. *HYDROLOGICAL PROCESSES*,
 20(2), 267–287.
- 547 Wondzell, S. M., LaNier, J., & Haggerty, R. (2009). Evaluation of alternative groundwater flow

- 548 models for simulating hyporheic exchange in a small mountain stream. *Journal of Hydrology*,
- 549 364(1–2), 142–151. Retrieved from http://www.sciencedirect.com/science/article/B6V6C-
- 550 4TSD9JR-3/2/d06e25e84c576a9525353b70ffa19121
- Zimmer, M. A., & McGlynn, B. L. (2017). Ephemeral and intermittent runoff generation processes in
 a low relief, highly weathered catchment. *Water Resources Research*, *53*(8), 7055–7077.
 https://doi.org/10.1002/2016WR019742
- 554

555 10 Supplementary Material

556 Supplementary material containing 5 tables and 30 figures accompanies this manuscript.

557 1 Data Availability Statement

- 558 Topographic data and stream discharge data are available from the H.J. Andrews Experimental
- 559 Forrest LTER database (<u>https://andrewsforest.oregonstate.edu/data</u>) as data sets GI010 and
- 560 HF004, respectively. Valley and stream geometries derived from the LiDAR are publicly available
- 561 (Ward, Zarnetske, et al., 2019). Model outputs including timeseries of flowing length, contiguous
- length, and flowing frequency for each modeled segment are archived in CUAHSI's HydroShare at:
- 563 http://www.hydroshare.org/resource/2f9643bb5d85436ba997a466ba8ed653