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Title: Modelling the effects of land cover and climate change on soil water partitioning in a boreal headwater catchment

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Keywords: boreal climate change; land cover change; climate-vegetation interactions; water balance; soil hydrology; Hydrus-1D

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Abstract: Climate and land cover are two major factors affecting the water fluxes and balance across spatiotemporal scales. These two factors and their impacts on hydrology are often interlinked. The quantification and differentiation of such impacts is important for developing sustainable land and water management strategies. Here, we calibrated the well-known Hydrus-1D model in a data-rich boreal headwater catchment in Scotland to assess the role of two dominant vegetation types (shrubs vs. trees) in regulating the soil water partitioning and balance. We also applied previously established climate projections for the area and replaced shrubs with trees to imitate current land use change proposals in the region, so as to quantify the potential impacts of climate and land cover changes on soil hydrology. Under tree cover, evapotranspiration and deep percolation to recharge groundwater was about 44% and 57% of annual precipitation, whilst they were about 10% lower and 9% higher respectively under shrub cover in this humid, low energy environment. Meanwhile, tree canopies intercepted 39% of annual precipitation in comparison to 23% by shrubs. Soils with shrub cover stored more water than tree cover. Land cover change was shown to have stronger impacts than projected climate change. With a complete replacement of shrubs with trees under future climate projections at this site, evapotranspiration is expected to increase by ~39% while percolation to decrease by 21% relative to the current level, more pronounced than the modest changes in the two components (<8%) with climate change only. The impacts would be particularly marked in warm seasons, which may result in water stress experienced by the vegetation. The findings provide an important evidence base for adaptive management strategies of future changes in low-energy humid environments, where vegetation growth is usually restricted by radiative energy and not water availability while few studies that quantify soil water partitioning exist.

Response to Reviewers:

Highlights

- Land cover and climate change impacts on soil water balance were modelled
- Vegetation effects were mainly reflected by evapotranspiration and interception
- Land cover change is expected to exert stronger impacts than climate change
- More marked impacts were projected to occur in warm seasons than cold seasons

1 Modelling the effects of land cover and climate change on soil water
2 partitioning in a boreal headwater catchment

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39 vegetation. The findings provide an important evidence base for adaptive
40 management strategies of future changes in low-energy humid environments, where
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44 water balance; soil hydrology; Hydrus-1D

45 **1. Introduction**

46 Changes in land surface hydrology reflect the combined effects of climate, vegetation
47 and soil (Rodriguez-Iturbe *et al.*, 2001; Li *et al.*, 2017). Climate, hydrology and
48 vegetation are intricately linked and the ecohydrological consequences of climate
49 change (CC) have been broadly discussed (Carey *et al.*, 2010; Tetzlaff *et al.*, 2013;
50 Xu *et al.*, 2013). For example, warming temperatures and increasing annual
51 precipitation (P) have resulted in an advanced vegetation green-up timing and
52 extended growing season in the northern hemisphere (Richardson *et al.*, 2013; Yang
53 *et al.*, 2015), and reduced snow accumulation with earlier melt (Knighton *et al.*, 2017).
54 Reduced summer rainfall and increased evapotranspiration (ET) also affect
55 streamflow generation (Déry and Wood, 2005; Deutscher *et al.*, 2016) and soil water
56 and groundwater storage (Barnett *et al.*, 2005; McNamara *et al.*, 2005; House *et al.*,
57 2016).

58 In addition to climate change, land cover change (LC) has been recognized as a key
59 factor that influences catchment hydrology (Zhang *et al.*, 2001; Li *et al.*, 2017). It is
60 estimated that vegetation covers ~70% of the global land surface (Dolman *et al.*,
61 2014), influencing water, carbon and energy exchanges driven by hydrological and
62 climatological factors (LeMone *et al.*, 2007). In particular, it has been estimated that
63 transpiration (T) contributes more than half the global terrestrial ET (Jasechko *et al.*,
64 2013), whilst precipitation interception (I) by the vegetation canopy can significantly
65 influence water redistribution (Carlyle-Moses and Gash, 2011; Soulsby *et al.*, 2017a).

66 Changes in land cover can have profound hydrological implications. For example, a
67 shift from grasses to trees would generally increase I and enhance ET (Brown *et al.*,
68 2005; Nunes *et al.*, 2011). Replacement of natural ecosystems by rain-fed agriculture
69 often results in increases in recharge and rising water tables (Allison *et al.*, 1990;
70 Scanlon *et al.*, 2005), while afforestation by deep rooted trees can reduce drainage and
71 lower water tables (Engel *et al.*, 2005). Many other studies (Favreau *et al.*, 2009) have
72 also demonstrated that LC can alter the catchment water balance significantly.

73 Climate change and vegetation development are interlinked and often coevolve
74 (Walther, 2010). Climate change impacts can be observed in the long term, for
75 instance, from precipitation and runoff data (Serreze *et al.*, 2000; Meng *et al.*, 2016),
76 whereas the impacts of land cover change can be expressed rapidly in runoff and
77 water chemistry responses (Séguis *et al.*, 2004; Guan *et al.*, 2013). In some cases,
78 climate change has been found to influence the hydrology of systems less
79 dramatically than land use/land cover change (Schilling *et al.*, 2010; Li *et al.*, 2017),
80 whilst others came to the opposite conclusion (Legesse *et al.*, 2003; Liu *et al.*, 2013).
81 Differentiation of their impacts on hydrological processes can help guide future
82 strategies to manage land and water in a more sustainable way.

83 The northern high-latitude regions are particularly sensitive to climate change (IPCC,
84 2014), though these changes will have significant spatial variability. It is estimated
85 that annual zonally averaged P increased by 7%–12% for latitudes of 30° N–85° N
86 over the 20th century (Dore, 2005). The figure for Canada was >10% on average over
87 a similar period (Mekis and Hogg, 1999), while over the United States it was 5%–10%
88 since 1900, most pronounced during warm seasons (Groisman *et al.*, 1999). The

89 spatial variation of climate change results in different impacts on local catchment
90 hydrology, especially in headwater catchments, which are important sources of stream
91 flow and groundwater recharge (Viviroli *et al.*, 2003). In many northern high latitudes
92 such as the Scottish Highlands, vegetation growth and productivity is usually
93 restricted by radiative energy and not water availability (Wang *et al.*, 2017a). The
94 natural vegetation over much of the Scottish Highlands would have been forests
95 dominated by Scots pine (*Pinus sylvestris*), but a long history of clearance, burning
96 and overgrazing has reduced forest cover dramatically (Steven and Carlisle, 1959).
97 With frequent rainfall, low radiation and high humidity, plants are usually not under
98 water stress during most of the year (Haria and Price, 2000). However, future
99 projections of intensified warming and decreased rainfall during growing seasons
100 (Gosling, 2012; Capell *et al.*, 2013), in addition to plans to increase Scots pine cover
101 to replace shrubs for conservation and biofuel objectives (Hrachowitz *et al.*, 2010),
102 may result in trees experiencing increased water stress in certain summer periods as
103 well as an increased annual *ET* and decreased water storage. Whilst this may have
104 advantages in terms of natural flood alleviation (Soulsby *et al.*, 2017b), it may also
105 reduce river flows to the detriment of in-stream ecology (Fabris *et al.*, 2017).

106 Numerous models have been developed to investigate soil water balance and its
107 interactions with climate and land cover changes (Romano, 2014; Ferguson *et al.*,
108 2016; Koch *et al.*, 2016). Among the most commonly used numerical solutions based
109 on the Richards equation for variably saturated water flow, the Hydrus-1D model has
110 been widely and successfully adopted for many cases ranging from laboratory
111 experiments to field study (Sutanto *et al.*, 2012; Ebel, 2013; Balugani *et al.*, 2017). In

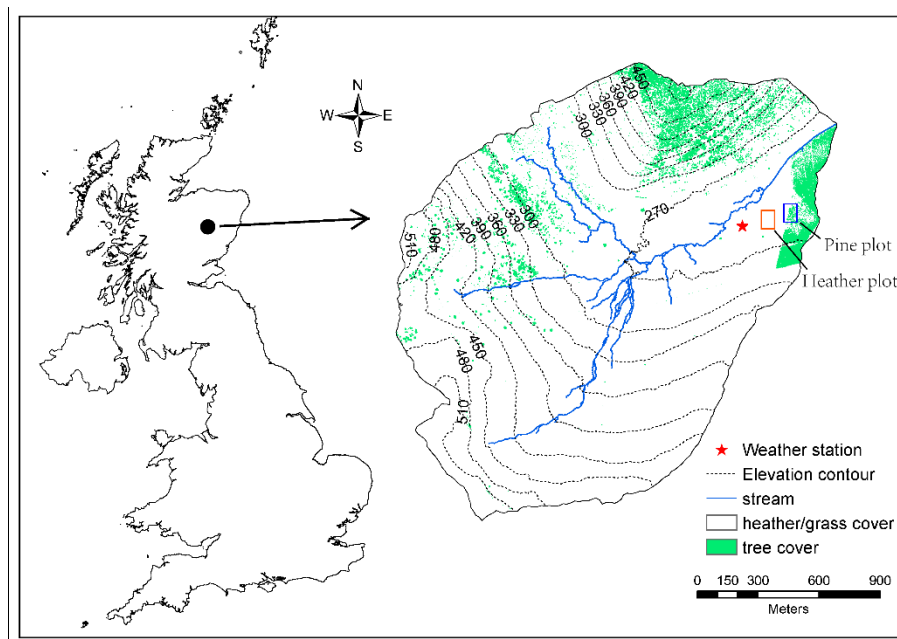
112 the most recent development, an interception module has been incorporated to
113 account for the role of vegetation in P redistribution (Šimůnek *et al.*, 2016).

114 In this study, we applied the Hydrus-1D model (the latest version 4.16) for podzolic
115 soils with two dominant vegetation covers (heather shrubs: *Ericaceae* vs. trees: *Pinus*
116 *sylvestris*) in two plots located in a Scottish headwater catchment (Bruntland Burn
117 catchment). The catchment's hydrology in terms of water transport, connectivity and
118 storage in the surface and subsurface, as well as runoff generation processes has been
119 extensively investigated using different advanced measurements and modeling
120 techniques such as stable water isotopes, geophysical surveys and tracer-aided models
121 (Tetzlaff *et al.*, 2014; Soulsby *et al.*, 2015, 2016b; van Huijgevoort *et al.*, 2016;
122 Benettin *et al.*, 2017). In the context of likely foreseeable CCs and LCs, the role of
123 vegetation in regulating the water balance in the unsaturated soils seems more
124 important (Geris *et al.*, 2015c) but is not yet fully understood (Tetzlaff *et al.*, 2015).
125 Therefore, the aims of this study are to quantitatively use a modelling approach to: (1)
126 investigate the effects of different vegetation covers on soil water balance components,
127 including I , ET , deep percolation (D) to groundwater recharge, and soil water storage
128 (S) in a boreal headwater catchment; and (2) examine and differentiate the impacts of
129 projected climate change and land cover change on the above soil water balance
130 components. The results will provide an evidence base and approach to guide
131 adaptive management in similar boreal sites, where such quantifications have not been
132 conducted before.

133 **2. Data and methods**

134 **2.1. Study site**

135 The Bruntland Burn catchment (3.2 km², 57.04°N, 3.13°W) is located in NE Scotland
136 (Figure 1), and described in detail elsewhere (Tetzlaff *et al.*, 2014; Ala-aho *et al.*,
137 2017). The climate is boreal oceanic. Based on the last decade of observations, mean
138 monthly maximum temperature is 19.4±1.3°C in July, and mean monthly minimum
139 temperature is -1.0±1.6°C in January. Mean annual *P* is around 1000 mm, relatively
140 evenly distributed throughout the year, but generally lower (~65 mm/month) in April-
141 July and higher (~105 mm/month) in October-February. Snow is generally <5% of
142 annual *P* and tends to lie for short periods (a few days to a few weeks) in January and
143 February and melts quickly. Annual mean potential evapotranspiration (*ET_p*) based on
144 the Penman-Monteith method (Allen *et al.*, 1998) is ~450 mm and annual runoff at
145 the catchment outlet is around 700 mm (Soulsby *et al.*, 2015).



146

147 Figure 1 The Bruntland Burn catchment and its location on the map of the United
148 Kingdom. The pine site was equipped with sap flow sensors and the heather site with
149 iButton sensors for transpiration estimation. Soil moisture probes were installed at
150 depths of 10, 20 and 40 cm in both the heather and pine plots. Numbers on contour
151 lines are elevation in m.a.s.l. Tree mapping in the figure is based on 1 m resolution
152 LiDAR data with the canopy height threshold set to 1.5 m.

153 Elevation in the catchment ranges from around 250 m.a.s.l at the valley bottom to
154 about 550 m on the ridge. Most of the underlying bedrock is granite, with Ca-rich and
155 Si-rich meta-sediments. Glacial drift deposits cover large parts of the catchment
156 (~70%) reaching up to 40 m deep in the valley bottom where this drift overlays the
157 bedrock (Soulsby *et al.*, 2007). In the valley bottom, the drift is comprised of a silty-
158 sand matrix with abundant larger clasts and has low permeability. In contrast, the
159 steeper hillslopes are veiled by shallower (~5 m deep), more permeable lateral
160 moraines and ice marginal deposits (Soulsby *et al.*, 2016b). Organic-rich soils
161 dominate the catchment, with large areas of deep peats (>1 m) in valley bottoms and
162 shallow peats (<0.5 m) on the lower hillslopes. On the steeper slopes, the dominant
163 soils covering ~60% of the catchment are podzols with a 0.1–0.2 m deep O horizon
164 on top. The dominant vegetation is heather (*Calluna vulgaris* and *Erica tetralix*)
165 shrubs with a canopy height of 0.3–0.6 m, distributed throughout the valley and
166 hillslopes. Trees, mostly Scots pine (*Pinus sylvestris*), cover about 10% of the
167 catchment, mainly in plantations near the outlet and natural forest on the south-facing
168 steeper slopes (Figure 1). Both heather and pine are evergreen vegetation that have
169 dense canopy. The majority of roots of heather and pine are present in the upper 0.15
170 and 0.3 m of the soils, respectively (Geris *et al.*, 2015a; Sprenger *et al.*, 2017).

171 **2.2.Measurements**

172 An automatic weather station (*Environmental Measurement Limited*, North Shields,
173 UK) was set up in the valley bottom of the catchment (Figure 1), continuously
174 recording 15-minute meteorological data, including temperature, relative humidity,
175 wind speed and direction, pressure, precipitation and net radiation and ground heat
176 flux. To estimate transpiration (T) from Scots pine, we installed 32 sets of sap flow
177 sensors (Thermal Dissipation Probes, *Dynamax Inc.* Huston, USA) on 10 trees at
178 trunk positions around 1.3 m above ground during 08/07-28/09/2015 (Wang *et al.*,
179 2017a). Transpiration of the plot was up-scaled by multiplying the sap flow rate per
180 sapwood area by sapwood area index (ratio of sapwood area to ground area) estimated
181 by the relationship between sapwood area (measured using wood cores e.g., Wang *et*
182 *al.*, (2016a)) and trunk diameter at 1.3 m height. Meanwhile, heather T between
183 31/07-31/10/2015 and 21/04-04/08/2016 was estimated using a model based on the
184 theory of maximum entropy production (MEP) (Wang and Bras, 2011), with canopy
185 temperature and humidity data collected every 15 minutes by 15 shielded iButton
186 sensors (DS1923 model, *Maxim Integrated*, USA). The sensors were fixed directly
187 over the heather canopy to minimize the influence of evaporation from interception on
188 rainy days, and were distributed in a 4 m by 8 m plot, which allows an estimation of
189 the spatial variability of water vapor distribution (Wang *et al.*, 2017b). Despite the
190 uncertainties related to the MEP model, transpiration was comparable to other
191 independent estimates (Wang *et al.*, 2017b). Soil water content was measured using
192 TDR probes (CS616, *Campbell Scientific, Inc.* USA) at both sites at depths of 10, 20
193 and 40 cm below the surface. All data were aggregated to daily values for further
194 comparisons and analysis. In addition, soil at different depths was sampled for

195 textural analysis. Leaf area index (LAI) was measured for pine (2.5 m²/m²) and
196 heather (1.7 m²/m²) using a plant canopy analyser LAI-2200C (*LI-COR*
197 *Environmental*, USA), and assumed to be constant throughout the year because both
198 heather and pine are evergreen and over 20 and 30 years old.

199 **2.3. Hydrus model configuration for heather and pine sites**

200 2.3.1. Governing equations in the Hydrus model

201 Simulation of the water fluxes and soil water storage dynamics were performed using
202 the Hydrus-1D model. We argue that the use of a 1D model for the unsaturated zone
203 is justified due to negligible lateral water flow along the small topographic gradients
204 at the study sites. Šimůnek *et al.*, (2013a) described the model structure and function
205 in detail. Only the dominant equations used in our study are summarised below.

206 One-dimensional variably-saturated water flow in the soil was modelled with the
207 modified Richards equation to account for the root water uptake by the sink term $S(h)$:

$$208 \quad \frac{\partial \theta}{\partial t} = \frac{\partial}{\partial x} \left[K(h, x) \left(\frac{\partial h}{\partial x} + 1 \right) \right] - S(h) \quad (1)$$

209 where θ is the volumetric water content (cm³/cm³), h is the water pressure head (cm),
210 t is time (day), x is the spatial coordinate (cm, positive upward), K is the unsaturated
211 hydraulic conductivity (cm/day). Unsaturated soil hydraulic properties $\theta(h)$ and $K(h)$
212 are given by *van Genuchten* (1980):

$$213 \quad \theta(h) = \begin{cases} \theta_r + \frac{\theta_s - \theta_r}{\left[1 + |\alpha h|^n\right]^{1-1/n}} & h < 0 \\ \theta_s & h \geq 0 \end{cases} \quad (2)$$

$$214 \quad K(h) = K_s S_e^l \left[1 - \left(1 - S_e^{1-1/n} \right)^{1-1/n} \right]^2 \quad (3)$$

215 where θ_r and θ_s are residual and saturated soil water content, respectively. α is the
 216 inverse of the air-entry value (or bubbling pressure). n is an empirical coefficient
 217 influencing the shape of hydraulic functions (dimensionless). K_s is saturated hydraulic
 218 conductivity (cm/day), l is the pore connectivity parameter prescribed as 0.5. S_e is
 219 effective saturation defined as the ratio of actual to maximum available soil water.

220 The sink term S (cm/day) for root water uptake calculation is defined in van
 221 Genuchten (1987):

$$222 \quad S(h) = f(h)\beta(x)T_p \quad (4)$$

$$223 \quad f(h) = \frac{1}{1 + \left(\frac{h}{h_{50}} \right)^p} \quad (5)$$

224 where $f(h)$ is a water stress response function, and $\beta(x)$ is the root distribution function
 225 (ranging 0-1) which linearly decreased in this study from the soil surface to 15 and 30
 226 cm below for heather and pine, respectively. h_{50} represents the water head at which
 227 the water extraction rate is reduced by 50%; p determines the $f(h)$ curve shape. These
 228 two parameters vary for vegetation and soils (Gribb *et al.*, 2009; Huang *et al.*, 2015).

229 In this study, we used -800 cm for h_{50} for both vegetation types; p was set to 3
 230 initially and tuned during calibration against transpiration estimates for heather and
 231 pine. Potential transpiration (T_p) was calculated from ET_p and soil cover fraction (f_s)
 232 in equation (6). ET_p was calculated separately for heather and pine following Dunn
 233 and Mackay (1995).

$$234 \quad T_p = f_s ET_p = (1 - e^{-\kappa \cdot LAI}) ET_p \quad (6)$$

235 κ is the light extinction coefficient, 0.56 for heather and 0.50 for pine according to
 236 Zhang *et al.*, (2014). Evaporation is calculated following Feddes *et al.*, (Feddes *et al.*,
 237 1974), that is, when the pressure head at the soil surface is higher than the minimum
 238 allowed pressure head (h_A), which is related to relative humidity and temperature, the
 239 actual evaporation is equal to the potential evaporation (E_p). Once the surface pressure
 240 head drops to h_A , the actual evaporation is decreased from E_p by solving equation (1).

241 In the Hydrus-1D version 4.16, precipitation interception by canopy is calculated
 242 following Kroes *et al.*, (2008):

$$243 \quad I = a \cdot LAI \cdot \left(1 - \frac{1}{1 + \frac{f_s \cdot P}{a \cdot LAI}} \right) \quad (7)$$

244 where P is precipitation (mm/day). The interception constant (a) was obtained by
 245 dividing the daily interception thresholds by LAI. Daily thresholds for heather and
 246 pine were 2.65 and 7.5 mm/day, respectively (Calder *et al.*, 1984; Haria and Price,
 247 2000).

248 2.3.2. Boundary conditions and soil hydraulic parameters

249 The upper boundary of the model was set as “atmospheric boundary (daily P and ET_p)
250 with surface runoff”, and the lower boundary was set as “free drainage” as the
251 groundwater table is usually deep on the hillslopes (>1 m) during the growing season.
252 Since root water uptake mainly occurs in the root-zone of the soils, we set up the
253 model for the upper 50 cm deep soil and configured the soil as two layers taking into
254 account the observed rooting depths: 0-15 cm and 15-50 cm for heather, and 0-30 cm
255 and 30-50 cm for pine.

256 The soil hydraulic parameters were initialized using the Rosetta (Schaap *et al.*, 2001)
257 estimates based on measured soil bulk density and texture (Geris *et al.*, 2015b;
258 Sprenger *et al.*, 2017), and then optimized with a Marquardt-Levenberg type
259 parameter estimation technique by inverse modelling (Šimůnek and Hopmans, 2002)
260 using θ observations at the depths of 10, 20 and 40 cm. The objective function was to
261 minimize the sum of squared difference between the observations and simulations at
262 each depth, and the goodness of fit was assessed by the coefficient of determination
263 (R^2) (Šimůnek *et al.*, 2013b). This method gives the mean as well as standard error of
264 parameter values, and has been tested in many studies and proved sufficient for soil
265 hydraulic parameter estimation (Schneider *et al.*, 2013; Li *et al.*, 2015; Bourgeois *et*
266 *al.*, 2016). Among the parameters in the soil hydraulic functions, the residual soil
267 moisture in equation (2) is considered not to influence model performance
268 significantly (Jacques *et al.*, 2002); therefore, it was fixed based on the Rosetta
269 estimates to reduce the number of parameters to calibrate.

270 2.3.3. Model validation

271 We chose two periods to calibrate the model covering both drying and wetting
272 conditions. Model performance was tested by comparing the simulations with direct
273 observations of soil moisture and also transpiration over two growing seasons using
274 data from sap flow (Wang *et al.*, 2017a) and iButton sensors (Wang *et al.*, 2017b) in
275 2015 and 2016. As measures of goodness of fit between simulations and observations
276 of both soil moisture and transpiration we applied R^2 and root mean square error
277 (*RMSE*) from linear regressions.

278 2.4. Climate and land cover change scenarios

279 We firstly ran a daily simulation for 05/2011-04/2016 to quantify the water
280 partitioning under two vegetation covers under current climate conditions. To
281 examine possible CC impacts, we then applied the projections for P and ET_p in the
282 area for the 2050s by Capell *et al.*, (2013), derived based on an average greenhouse
283 gas emission scenario (IPCC scenario A1B) in the UK Climate Projection 2009
284 (UKCP09) network, which downscaled Regional Climate Model predictions to 5 km
285 resolution. Specifically, compared to the current condition, P is projected to decrease
286 by 10% in April-October while it is projected to increase by 10% in November-March;
287 ET_p is projected to increase by 15% in all months because of warming temperatures
288 throughout the year. Lastly, to assess the possible LC impacts within a modelling
289 experiment, we assumed a 100% change from heather to pine running the simulations
290 with pine parameters in place of heather for the heather soil profile. The rationale for
291 this was the historic and current policy drive to increase tree cover in the Scottish
292 Highlands for conservation, biomass production, natural flood management, and

293 mitigating rising temperatures in streams for important species like Atlantic salmon
 294 (Tudor *et al.*, 1999; Hrachowitz *et al.*, 2010). Monthly averages of the water balance
 295 components (equation 8) within the soil columns over the 5-year period were
 296 compared to investigate the impacts of LC under future climate scenarios.

$$297 \qquad \qquad \qquad \Delta S = P - ET - D \qquad \qquad \qquad (8)$$

298 Where ΔS is the water storage change in the 50 cm deep soil columns, D is the deep
 299 percolation leaving the 50 cm soil columns. Lateral fluxes are neglected due to the
 300 relatively flat terrain conditions. To examine the significance of difference for each
 301 component among the three scenarios (current, climate, climate plus land cover
 302 change), we, firstly, performed an analysis of variance (ANOVA) and then a post hoc
 303 Tukey test with the significance level of 0.05. The data used for the test were yearly
 304 data from 2011 to 2016.

305 **3. Results**

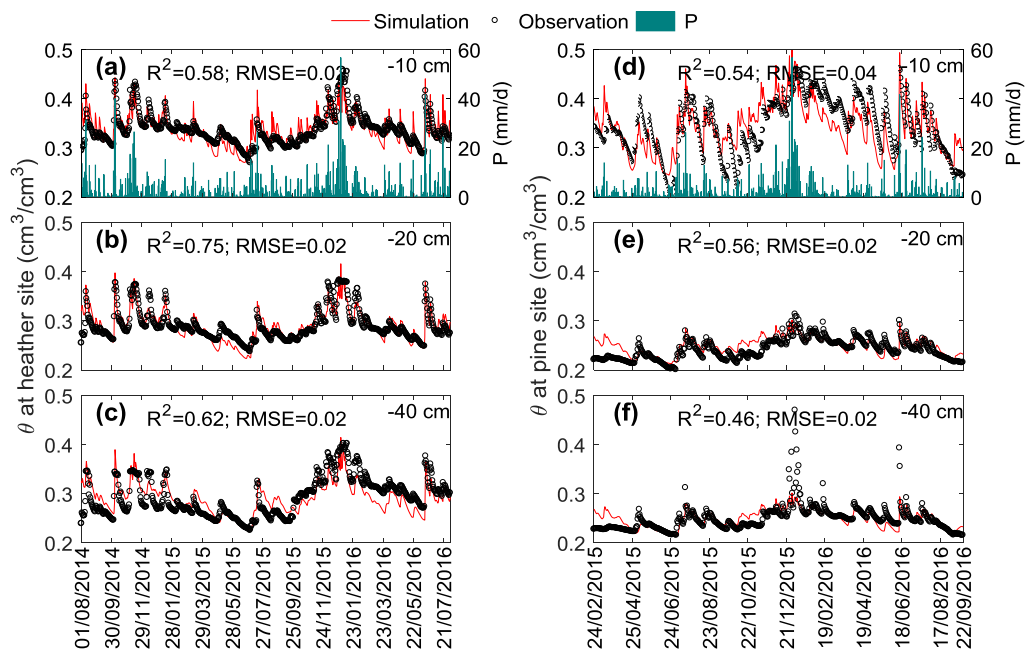
306 **3.1. Model test against soil moisture and transpiration**

307 Table 1 Calibrated mean soil hydraulic parameters in equation 2-3 for both sites. In
 308 brackets is standard error of each parameter. θ_s is saturated soil water content, K_s is
 309 saturated hydraulic conductivity, α is the inverse of the air-entry value (or bubbling
 310 pressure), and n is an empirical coefficient influencing the shape of hydraulic functions.

Site	Depth (cm)	θ_s (cm ³ /cm ³)	α (1/cm)	n (-)	K_s (cm/d)
Heather	0-10	0.5188 (0.0100)	0.0874 (0.0087)	1.1182 (0.0069)	195.02 (40.64)
	10-50	0.4965 (0.0098)	0.0386 (0.0035)	1.3090 (0.0089)	359.75 (62.19)
Pine	0-30	0.5736 (0.0154)	0.0706 (0.0070)	1.2088 (0.0077)	327.38 (72.32)
	30-50	0.3119 (0.0050)	0.0257 (0.0022)	1.1305 (0.0083)	308.33 (42.30)

311

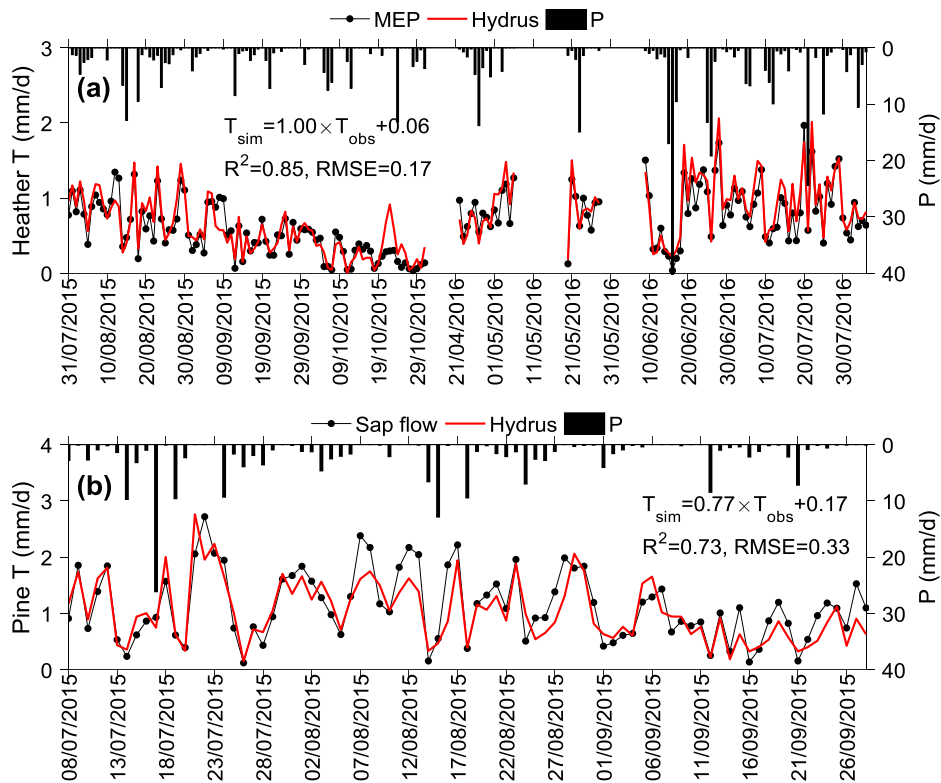
312 The parameters of the soil hydraulic functions were calibrated by inverse solution of
 313 the soil water content at 10, 20 and 40 cm depths. Optimized mean parameter values
 314 as well as the standard deviation are given in Table 1. The textures of the soils under
 315 heather and pine cover had no distinct difference (Sprenger *et al.*, 2017), thus, the
 316 calibrated parameters of the soil hydraulic functions were mostly similar, except θ_s ,
 317 and K_s . In general, the top layer had higher θ_s than the lower layer consistent with a
 318 higher organic matter content near the surface; also average K_s was higher at the pine
 319 site than the heather site, consistent with likely greater root biomass in the forest. The
 320 small standard deviations of the parameter values indicate that the inverse solution
 321 gave generally stable estimates with small uncertainties.



322

323 Figure 2 Comparison of simulated and measured soil water content (θ) over the
 324 calibration periods. Plots *a-c* are for heather site at the depths of 10, 20 and 40 cm,
 325 and plots *d-f* are for pine site at the same depths. P is precipitation. R^2 is coefficient of
 326 determination, and $RMSE$ is the root mean square error, with the unit of cm^3/cm^3 .

327 Model performance was firstly tested against soil moisture (Figure 2). Simulations
328 generally agreed well with observations in terms of both the temporal dynamics
329 reflected by R^2 (0.58-0.75 for the heather site and 0.46-0.56 for the pine site), and
330 water content reflected by $RMSE$ ($0.02 \text{ cm}^3/\text{cm}^3$ at all three depths for heather, and
331 $0.02 \text{ cm}^3/\text{cm}^3$ at 20 and 40 cm depths and $0.04 \text{ cm}^3/\text{cm}^3$ at 10 cm depth for pine). It
332 was notable that the forest top soil showed much greater variability in moisture
333 content than the heather site, probably due to more root water uptake by pines and
334 faster drainage through the shallower organic layer to dry the soils. It is also
335 noticeable that soil moisture contents were lower and less variable at the forest site
336 than the heather site, reflecting the much coarser subsoil characteristics (Dick *et al.*,
337 2017). Whilst at the heather site there was a tendency to slightly overestimate
338 moisture content in wet periods, for the forest site the model over-anticipated the wet-
339 up following summer 2015 and exaggerated the dry-down following very large
340 rainfall inputs in December 2015/January 2016. However, the over/under-estimations
341 did not consistently occur throughout the entire calibration periods. The soils at 40 cm
342 depth were similar to and slightly drier than at 20 cm depth indicating a generally
343 free-draining characteristic and a deep groundwater table (at least $>0.5 \text{ m}$) at both
344 sites, though in the December 2015 event, very high water content was briefly
345 observed in the subsoil of the forest which was not captured by the model.



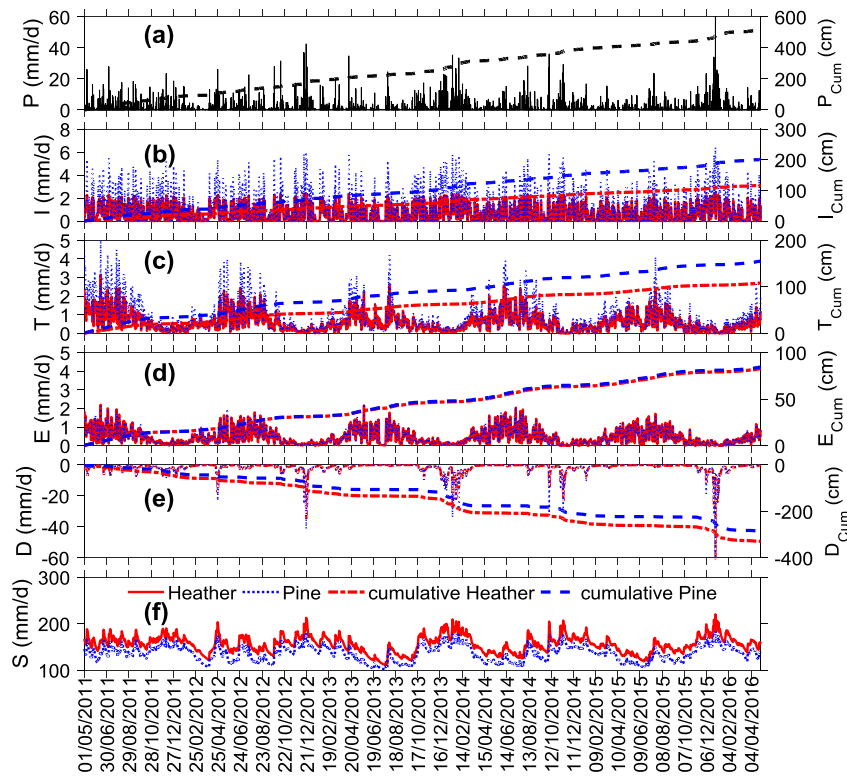
346

347 Figure 3 Comparison of transpiration (T) simulated by Hydrus and estimated from
 348 independent methods for (a) heather and (b) pine site. T at heather site was derived
 349 using the MEP method, and measured at pine site using the sap flow technique. $RMSE$
 350 is the root mean square error, in the unit of mm/d. Equations in the figure are linear
 351 regressions between Hydrus simulations and MEP-derived and sap flow measured T .

352 In addition to soil water content, model performance for transpiration simulations
 353 (Figure 3) was also tested against direct measurement or independent estimations
 354 from data driven models (i.e. measured via sap flow for Scots pine; derived from the
 355 MEP method for heather). Again, the Hydrus model showed good agreement with the
 356 T estimates for both vegetation types, capturing the dynamics well with a high R^2 and
 357 giving similar absolute fluxes shown with the low $RMSE$. There was a general
 358 tendency for the model to occasionally underestimate the observed values on some
 359 days but again this was not systematic. Noticeably, heather T (peak value ~ 2 mm/d)
 360 was lower than pine T (peak ~ 3 mm/d). For the period from summer towards autumn,
 361 there was a decreasing tendency of T from both heather and pine in response to

362 declining solar radiation (Wang *et al.*, 2017a). Overall, the good agreements between
 363 Hydrus simulations and observations of soil water content and independent estimates
 364 of transpiration increased confidence that the model was able to represent the soil
 365 water flow and plant water uptake processes reasonably well.

366 **3.2.Simulations of water balance components under two vegetation covers**



367

368 Figure 4 Comparison of key water balance components between heather and pine sites.
 369 (a) P : precipitation, (b) I : total intercepted precipitation by canopy, (c) T : transpiration,
 370 (d) E : evaporation, (e) D : deep percolation leaving the 50cm deep soils, and (f) S :
 371 total water storage in the 50cm soil profile. Right axes in the plots (a-e) show the
 372 cumulative values of the corresponding variables.

373 To compare the influence of vegetation cover on water partitioning, daily as well as
 374 cumulative values of P and simulated water balance components are given in Figure 4
 375 over a 5-year period from 05/2011 to 04/2016. P was relatively evenly distributed
 376 throughout the year, though slightly higher in winter than in summer. Similar to P ,

377 there was no strong seasonality in the total intercepted precipitation (*I*). Possibly
378 unsurprising, Scots pine canopy intercepted more precipitation than the heather
379 canopy (1.7-fold). This corroborates recent empirical work that has suggested
380 interception losses of ~40% in the forest (Soulsby *et al.*, 2017a) and 22% from the
381 heather (Wang *et al.*, 2017b). Transpiration was also higher from Scots pine than from
382 heather shrubs, showing the 5-year accumulation of ~1500 mm from pine compared
383 to ~1000 mm from heather; whereas evaporation (*E*) was similar at both sites (5-year
384 accumulation of ~760 mm), and significantly lower than transpiration. *ET* was high
385 between April-October and low during the rest of the year. Deep percolation to
386 recharge deeper soils (>50 cm) at the heather site was more than the pine site. Largest
387 percolation occurred in winter such as January 2016 during large rain events. In
388 summer, when *P* was low and *ET* was high, percolation effectively ceased for
389 prolonged periods. Soil water storage (*S*), defined as the total amount of water that is
390 stored in the soil profile and calculated as the available soil water storage capacity
391 multiplied by the soil depth, was high (nearly 200 mm) in winter and low (~100 mm)
392 in summer, which inversely reflected *ET* and corresponded positively to rainfall. On
393 average, *S* was around 159 mm at the heather site, and ~20 mm less at the pine site.

394 Table 2 Average annual water balance components in the upper 50 cm soils under
395 heather and pine covers from 05/2011 to 04/2016. Percentage in brackets shows
396 the mean (\pm standard deviation) proportion of annual precipitation for each
397 relevant component. *P*-precipitation, *ET*-evapotranspiration, and *D*-deep
398 percolation. ΔS is the water storage change which was calculated as *P-ET-D*.

Site	<i>P</i> (mm)	<i>ET</i> (mm)	<i>D</i> (mm)	ΔS (mm)
Heather	1039.3 \pm 86.4	355.0 \pm 31.2	693.9 \pm 69.9	-9.6 \pm 16.7
		(34.2 \pm 3.0%)	(66.8 \pm 6.7%)	(-0.9 \pm 1.6%)
Pine		459.1 \pm 60.1	596.4 \pm 69.5	-16.3 \pm 18.1
		(44.2 \pm 5.8%)	(57.4 \pm 6.7%)	(-1.6 \pm 1.7%)

399

400 The average annual amounts of the water balance components under different
 401 vegetation types over the 5-year period are given in Table 2. The annual precipitation
 402 was 1039 mm, and more than half of it drained below the top 50 cm soils to recharge
 403 deeper soil water or groundwater. In contrast, around $34\pm 3\%$ of annual precipitation
 404 was lost through *ET* from the heather site, and 10% more from the pine site. The
 405 average annual water storage change in the soils was only a small portion of
 406 precipitation. In addition, the interception calculations showed that heather
 407 intercepted $23\pm 2\%$ of annual *P* whereas pine intercepted $39\pm 3\%$. Based on the results,
 408 we can conclude that at our site vegetation effects on soil water balance are mostly
 409 reflected in *ET* and net precipitation, which consequently changes the recharge to
 410 deeper soils.

411 3.3.Impacts of land cover and climate changes on the soil water balance

412 Table 3 Changes in annual water balance components at the heather and pine sites
 413 under scenarios of climate and land cover changes relative to the current condition.
 414 Percentages show the average (\pm standard deviation) projected increase (positive) or
 415 decrease (negative) in each component. *P*-precipitation, *ET*-evapotranspiration, and
 416 *D*-deep percolation. ΔS is the water storage change calculated as the residual of *P*, *ET*,
 417 and *D*. Heather* means running the model with soil properties of the heather site but
 418 vegetation parameters of Scots pine. ** indicates that the Tukey test *p* value was <0.05 .

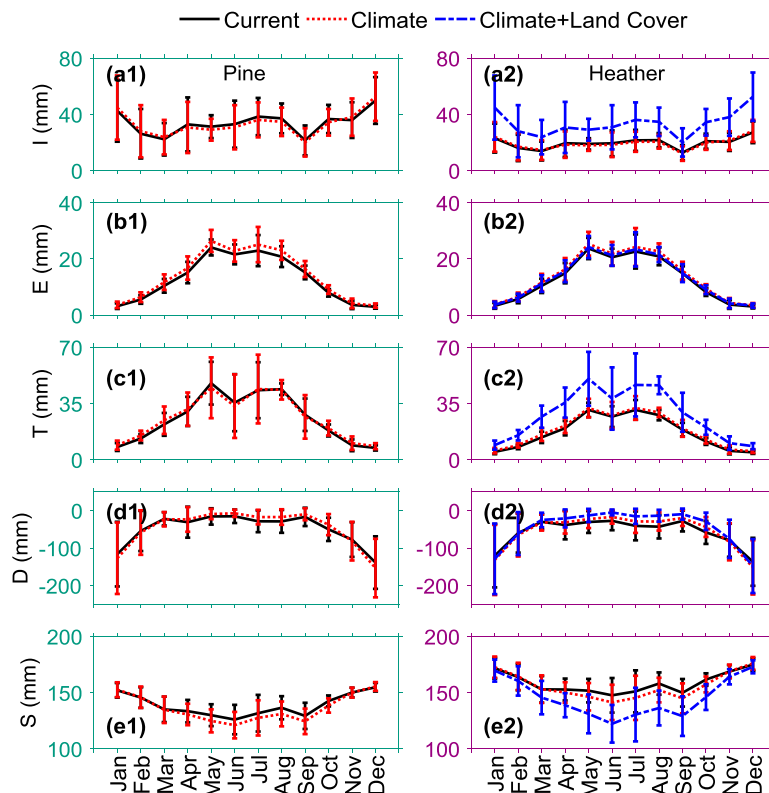
Site	Scenario	<i>P</i>	<i>ET</i>	<i>D</i>	ΔS
Heather	Climate change		$7.6\pm 3.5\%$	$-5.2\pm 6.7\%$	$13.1\pm 41.7\%$
Pine			$4.5\pm 5.2\%$	$-4.9\pm 9.7\%$	$11.6\pm 21.2\%$
	Climate change +	$-1.0\pm 3.2\%$		-	
Heather*	land cover change		$39.5\pm 15.1\%$ **	$21.3\pm 12.0\%$	$33.3\pm 92.7\%$ **

419 Climate trends and projections indicate a warming environment in northern high
 420 latitudes, changes in seasonal temperature distributions as well as altered precipitation
 421 regimes are expected to exert impacts on water balance. These changes will be

422 spatially variable and need to be clearly quantified locally. Here, we show firstly the
423 corresponding changes in water balance components under projected climate change
424 for the 2050s (annual P and ET_p) in comparison to the current condition in Table 3.
425 These were derived from running the calibrated Hydrus model for the two sites with
426 climatic data modified according to the IPCC average greenhouse gas emission
427 scenario for the region. As annual P is only expected to decrease slightly by the 2050s
428 (and the change was not statistically significant with the Tukey test $p=0.93$), ET was
429 modelled to increase by $7.6\pm 3.5\%$ at the heather site and $4.5\pm 5.2\%$ at the pine site
430 (both with a $p=0.62$). The smaller increase in ET at the pine site may indicate that
431 more constraints from root-zone soil water availability would be expected by the trees.
432 As a result of reduced P and enhanced ET , deep percolation was projected to decrease
433 by 5.2% ($p=0.95$) and 4.9% ($p=0.79$) for heather and pine, respectively. It is notable
434 that D remained the largest water component in the soils. The dynamics of ΔS were
435 modelled as increasing at both sites, by 13.1% ($p=0.78$) and 11.6% ($p=0.41$) for
436 heather and pine with large variability. The absolute value of annual ΔS for heather
437 and pine under future climate was small, amounting to -10.9 and -18.2 mm/year,
438 respectively. In addition, with the reduced P , intercepted precipitation was modelled
439 to decrease by $1.2\pm 1.3\%$ ($p=0.99$) from pine and $1.0\pm 0.8\%$ ($p=0.90$) from heather.
440 Overall, climate change would have impacts on the soil water balance components but
441 the impacts were tested not statistically significant.

442 As there are policy drivers to promote changes in land cover from heather to pine to
443 mitigate future CC impacts, we also ran the simulations for the heather-covered soils
444 but replaced the relevant heather vegetation parameters by pine values (i.e. root
445 distribution, LAI, light extinction coefficient, interception capacity and constant a ,

446 and ET_p), and calculated the changes in water budgets in Table 3. Compared to the
 447 current conditions, when considering both CC and LC, we found that ET would
 448 increase dramatically by $39.5 \pm 15.1\%$ ($p < 0.001$), and deep percolation would decrease
 449 by $21.3 \pm 12.0\%$ ($p = 0.42$). The shift of land cover types would mainly result in changes
 450 of annual ET and percolation, and consequently, ΔS would increase by 33.3% from -
 451 9.6 to -12.8 mm/year ($p = 0.01$), but with large variability. Moreover, shifting from
 452 heather to pine under future climate was estimated to increase canopy intercepted
 453 water significantly by $71.2 \pm 5.8\%$ ($p < 0.001$) compared to the current heather
 454 dominated conditions. Therefore, the complete replacement of heather with pine
 455 under future climate is expected to alter evapotranspiration, canopy interception and
 456 soil water storage most significantly.



457

458 Figure 5 (a1-e1) Climatological mean monthly soil water balance components for
459 current condition and projected climate change at the pine site; and (a2-e2) mean
460 monthly soil water balance components for current condition, projected climate
461 change, and climate change plus land cover change at the heather site. Error bars
462 represent standard deviation of the monthly values for each variable in 5 years. *I*-
463 canopy interception of precipitation, *E*-evaporation, *T*-transpiration, *D*-deep
464 percolation, *S*-total water storage in the 50 cm soils.

465 To examine the seasonality of hydrological variables as well as the degree of probable
466 impacts of CC and LC, we also calculated the monthly average of the water balance
467 components over the 5-year period, and results are given in Figure 5. Canopy
468 interception of precipitation showed similar dynamics to *P*, i.e. generally high in
469 January and December and low in February and March. *ET*, *D* and *S* showed much
470 stronger seasonal variations than interception. *ET* was higher in April-September than
471 in other months, whilst *D* and *S* were smaller in summer than winter. Large variability
472 corresponds to seasons with a large amount of water components. It is worth
473 mentioning that changes in evaporation in all cases were projected to be negligible,
474 probably due to the limited radiation below the dense canopies of both vegetation.
475 Therefore, the increase in *ET* would be primarily contributed by transpiration. Climate
476 change would have relatively modest impacts on all examined water balance
477 components. In contrast, land cover change of increased forest cover has a stronger
478 potential influence. The simulations show that shifting from heather to Scots pine
479 would result in more plant interception and transpiration, lower soil water storage and
480 reduced recharge especially in growing seasons.

481 **4. Discussion**

482 **4.1.Role of vegetation in soil water partitioning**

483 The northern high latitudes are experiencing land cover changes (Kelly and Goulden,
484 2008; Bi *et al.*, 2013), which are likely to increase through time and have the potential
485 to exert strong but variable influence on the hydrology of this region (Tetzlaff *et al.*,
486 2015). Different land cover types have contrasting effects on water partitioning.
487 Generally, trees transpire more water, and lose more water from precipitation
488 interception than grass and shrubs (Zhang *et al.*, 2001; Kroes *et al.*, 2008), and thus,
489 runoff generation is usually higher in areas with more grass/shrub cover than forest
490 cover (Sterling *et al.*, 2012). Therefore, manipulating vegetation cover has potential
491 for land and water conservation, e.g., to buffer soil erosion or losses of water (Nunes
492 *et al.*, 2011) through high *ET* or rapid runoff (Van Der Linden *et al.*, 2003). Results
493 from our work show that the role of vegetation in partitioning water is primarily
494 mediated by interception and subsequent transpiration, which eventually results in
495 differing summer water use and soil moisture storage under two distinct vegetation
496 covers as also shown by Vivoni *et al.*, (2008). These results are consistent with
497 previous forest hydrology studies in the UK (Dunn and Mackay, 1995; Haria and
498 Price, 2000; Sprenger *et al.*, 2017) in that *ET* from Scots pine was much higher than
499 heather, reflecting the greater water use through transpiration and higher interception
500 capacity of Scots pine due to the higher LAI (Haria and Price, 2000). The surface
501 runoff was negligible in our simulations because the two sites under survey are on
502 gentle slopes with well-vegetated, free-draining soil profiles (Tetzlaff *et al.*, 2007). In
503 addition, rainfall intensities are usually low, though some empirical evidence
504 suggested overland flow may have occurred after the largest rainfall events in winter

505 2015/2016 (Figure 4). However, the frequency of such events is low and the annual
506 average amount of surface runoff was trivial compared to other water balance
507 components (Ala-aho *et al.*, 2017). The resulting percolation of water to groundwater
508 recharge is consistent with the role of deeper subsurface flows being the dominant
509 contribution to catchment runoff generation from the podzolic soils (Blumstock *et al.*,
510 2016; Soulsby *et al.*, 2016b). Note that the soil textures show limited differences at
511 both sites, therefore, the difference in water balance components is primarily resulted
512 from vegetation cover effects.

513 In our low-energy humid environment, *ET* is primarily driven by solar radiation
514 (Tetzlaff *et al.*, 2015; Wang *et al.*, 2017a), unlike in semi-arid areas where *ET* is also
515 significantly controlled by soil water availability (Wang *et al.*, 2016b). Global land
516 surface *ET* is dominated by *T*, and the annual average *T/ET* ratio given in Miralles *et*
517 *al.*, (2011) was 0.80, in the upper range reported by others with a lower average ratio
518 of 0.67 (Jasechko *et al.*, 2013; Schlesinger and Jasechko, 2014). In our study, the
519 annual average *T/ET* ratio was 0.57 at the heather site and 0.67 at the pine site, within
520 the range previously reported. Meanwhile, the proportion of *ET* in *P* at the heather
521 and pine sites was ~34% and 44%, lower than the average value of 58% for Europe
522 (Miralles *et al.*, 2011), but comparable to similar work on Scots pine plots in Belgium
523 (Verstraeten *et al.*, 2005). Noticeably, the biggest difference in water partitioning in
524 this study compared to those in drier regions (Li and Shao, 2014) lies in the high
525 recharge to deeper soil horizons and groundwater. We found that more than half of
526 the precipitation drained to the deeper (>0.5 m) soils. The abundant *P* and *D* ensures
527 that groundwater storage is recharged in the autumn and winter to sustain baseflows
528 (Figure 4-5), whilst soil moisture deficits are usually negligible at the start of the

529 growing season and water is available for vegetation uptake (Tetzlaff *et al.*, 2007;
530 Soulsby *et al.*, 2016a). The downslope movement of groundwater storage from the
531 hillslopes also sustains saturated conditions in the valley-bottom riparian zones where
532 most of total water storage in the catchment resides (Soulsby *et al.*, 2016b).

533 **4.2. The impacts of climate change and land cover change**

534 Climate change and ecosystem composition shifts as well as their impacts on
535 hydrology often occur concurrently (Kelly and Goulden, 2008). However, the exact
536 influences are variable and contrasting in different regions (Legesse *et al.*, 2003; Li *et al.*,
537 2017). Thus, understanding both the integrated and separate effects of CC and LC
538 on local water balance components is of particular importance to project likely future
539 changes and adapt to them (Ray *et al.*, 2010). In our study, we showed that with the
540 UKCP09 projected changes in P and ET_p by the 2050s, the most pronounced change
541 would be transpiration and interception loss from both vegetation types. Similar
542 findings have been reported by others across different climates and landscapes (Ivits
543 *et al.*, 2014; House *et al.*, 2016). Currently, the vegetation in our catchment rarely
544 experiences water stress during growing seasons (Wang *et al.*, 2017a); however, with
545 the projected drier and warmer summers in that region, plants are more likely to
546 experience water shortages, partly because they are mostly established on shallow
547 podzolic soils on hillslopes with high percolation rates recharging groundwater which
548 is usually below the rooting zones (Geris *et al.*, 2015a), and also because more soil
549 water storage would be lost through ET than it is now (Table 3). Thus, with a
550 reduction in S and an increase in ET , the water available for vegetation, groundwater

551 recharge and baseflow maintenance would be decreased. However, it should be
552 stressed that these changes are relatively modest.

553 The simulations suggest that potential LCs have a stronger impact on the soil water
554 balance than CC at this particular site. The most pronounced influence was projected
555 to be an increased vegetation water use with a complete shift from heather shrubs to
556 trees. The results are consistent with similar studies such as Beringer *et al.*, (2005)
557 who showed an increase in *ET* with the shift from tundra shrubs to trees in Alaska,
558 and Scanlon *et al.*, (2005) who showed replacement of rangeland by agriculture
559 resulted in less *ET* and more recharge in southwestern US. Looking into the 5-year
560 average monthly values, we found the changes in water balance would primarily
561 occur in warm and dry seasons (April-October) rather than in cold and wet seasons
562 (November-March) though an increase in temperature (and ET_p) is expected to occur
563 throughout the year (Capell *et al.*, 2012; Gosling, 2012). It should be noted that in this
564 study we only considered the change in amount but not the change in intensity of
565 precipitation which could influence the interception as well as recharge (Love *et al.*,
566 2010; Zhang *et al.*, 2016). Current scenarios suggest an increase in summer rainfall
567 intensities, which could reduce interception losses that tend to be high when rainfall
568 intensity is low (Soulsby *et al.*, 2017a). Moreover, the LC scenario was also set to a
569 complete status by the 2050s; the transitional period from clearance of heather to
570 maturity of Scots pine was not considered in the simulations. During this period, the
571 quantitative effects of vegetation on soil hydrology would likely be different.
572 Nonetheless, the results are instructive and can help with the understanding of both
573 climate and land cover impacts on soil hydrology in the catchment and other boreal
574 environments in the long term.

575 **5. Conclusions**

576 We applied the Hydrus-1D model to two representative soil-vegetation units in a
577 boreal headwater environment to quantify the soil water balance and to investigate the
578 potential climate change and land cover impacts. Our work showed that the role of
579 vegetation in partitioning water is primarily mediated by interception and subsequent
580 transpiration, which eventually results in differing summer water use and soil
581 moisture storage, as well as the recharge to deeper soils. Transpiration and
582 interception both are higher from Scots pine than heather shrubs. Soil water storage
583 shows the reverse pattern, higher from heather, though differences are seasonally
584 focused in the summer when evapotranspiration is highest. With climate projections
585 and proposed land cover changes, the majority of changes to soil water balance are
586 projected to occur in the growing seasons. Potential land cover changes seem to have
587 stronger impacts than current climate change projections on local water balances at
588 the study site. This study applies integrated field data and modelling approaches to
589 enhance our understanding of potential land cover and climate change impacts on the
590 soil hydrology in a boreal, humid environment, and as such provides an evidence base
591 for policy development and land use planning to protect water supplies and ecosystem
592 services.

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- 599 Ala-aho P, Soulsby C, Wang H, Tetzlaff D. 2017. Integrated surface-subsurface
600 model to investigate the role of groundwater in headwater catchment runoff
601 generation: A minimalist approach to parameterisation. *Journal of Hydrology*
602 **547**: 664–677 DOI: 10.1016/j.jhydrol.2017.02.023
- 603 Allen RG, Pereira LS, Raes D, Smith M. 1998. Crop evapotranspiration - Guidelines
604 for computing crop water requirements - FAO Irrigation and drainage paper 56.
605 *Irrigation and Drainage*: 1–15 DOI: 10.1016/j.eja.2010.12.001
- 606 Allison GB, Cook PG, Barnett SR, Walker GR, Jolly ID, Hughes MW. 1990. Land
607 clearance and river salinisation in the western Murray Basin, Australia. *Journal*
608 *of Hydrology* **119** (1–4): 1–20 DOI: 10.1016/0022-1694(90)90030-2
- 609 Balugani E, Lubczynski MW, Reyes-Acosta L, van der Tol C, Francés AP, Metselaar
610 K. 2017. Groundwater and unsaturated zone evaporation and transpiration in a
611 semi-arid open woodland. *Journal of Hydrology* **547**: 54–66 DOI:
612 10.1016/j.jhydrol.2017.01.042
- 613 Barnett TP, Adam JC, Lettenmaier DP. 2005. Potential impacts of a warming climate
614 on water availability in snow-dominated regions. *Nature* **438** (7066): 303–9 DOI:
615 10.1038/nature04141
- 616 Benettin P, Soulsby C, Birkel C, Tetzlaff D, Botter G, Rinaldo A. 2017. Using SAS
617 functions and high-resolution isotope data to unravel travel time distributions in
618 headwater catchments. *Water Resources Research* **53** (3): 1864–1878 DOI:
619 10.1002/2016WR020117
- 620 Beringer J, Chapin FS, Thompson CC, McGuire AD. 2005. Surface energy exchanges
621 along a tundra-forest transition and feedbacks to climate. *Agricultural and Forest*
622 *Meteorology* **131** (3–4): 143–161 DOI: 10.1016/j.agrformet.2005.05.006
- 623 Bi J, Xu L, Samanta A, Zhu Z, Myneni R. 2013. Divergent Arctic-Boreal vegetation
624 changes between North America and Eurasia over the past 30 years. *Remote*
625 *Sensing* **5** (5): 2093–2112 DOI: 10.3390/rs5052093
- 626 Blumstock M, Tetzlaff D, Dick JJ, Nuetzmann G, Soulsby C. 2016. Spatial
627 organization of groundwater dynamics and streamflow response from different
628 hydrogeological units in a montane catchment. *Hydrological Processes* **30** (21):
629 3735–3753 DOI: 10.1002/hyp.10848
- 630 Bourgeois O Le, Bouvier C, Brunet P, Ayrat P-A. 2016. Inverse modeling of soil
631 water content to estimate the hydraulic properties of a shallow soil and the
632 associated weathered bedrock. *Journal of Hydrology* **541**, Part: 116–126 DOI:
633 <http://dx.doi.org/10.1016/j.jhydrol.2016.01.067>
- 634 Brown AE, Zhang L, McMahon TA, Western AW, Vertessy RA. 2005. A review of
635 paired catchment studies for determining changes in water yield resulting from
636 alterations in vegetation. *Journal of Hydrology* **310** (1–4): 28–61 DOI:
637 10.1016/j.jhydrol.2004.12.010
- 638 Calder IR, Hall RL, Harding RJ, Wright IR. 1984. The Use of a Wet-Surface
639 Weighing Lysimeter System in Rainfall Interception Studies of Heather (*Calluna*
640 *vulgaris*). *Journal of Climate and Applied Meteorology* **23** (3): 461–473 DOI:
641 [Doi 10.1175/1520-0450\(1984\)023<0461:Tuoaws>2.0.Co;2](https://doi.org/10.1175/1520-0450(1984)023<0461:Tuoaws>2.0.Co;2)
- 642 Capell R, Tetzlaff D, Hartley AJ, Soulsby C. 2012. Linking metrics of hydrological
643 function and transit times to landscape controls in a heterogeneous mesoscale
644 catchment. *Hydrological Processes* **26** (3): 405–420 DOI: 10.1002/hyp.8139

645 Capell R, Tetzlaff D, Soulsby C. 2013. Will catchment characteristics moderate the
646 projected effects of climate change on flow regimes in the Scottish Highlands?
647 *Hydrological Processes* **27** (5): 687–699 DOI: 10.1002/hyp.9626

648 Carey SK, Tetzlaff D, Seibert J, Soulsby C, Buttle J, Laudon H, McDonnell J,
649 McGuire K, Caissie D, Shanley J, et al. 2010. Inter-comparison of hydro-climatic
650 regimes across northern catchments: synchronicity, resistance and resilience.
651 *Hydrological Processes* **24** (24): 3591–3602 DOI: 10.1002/hyp.7880

652 Carlyle-Moses DE, Gash JHC. 2011. Rainfall interception loss by forest canopies.
653 *Forest Hydrology and Biochemistry: Synthesis of Past Research and Future
654 Directions* **216**: 407–424 DOI: 10.1007/978-94-007-1363-5_20

655 Déry SJ, Wood EF. 2005. Decreasing river discharge in northern Canada.
656 *Geophysical Research Letters* **32** (10): 1–4 DOI: 10.1029/2005GL022845

657 Deutscher J, Kupec P, Dundek P, Holík L, Machala M, Urban J. 2016. Diurnal
658 dynamics of streamflow in an upland forested micro-watershed during short
659 precipitation-free periods is altered by tree sap flow. *Hydrological Processes* **30**
660 (13): 2042–2049 DOI: 10.1002/hyp.10771

661 Dick JJ, Tetzlaff D, Bradford J, Soulsby C. 2017. Using repeat electrical resistivity
662 surveys to assess heterogeneity in soil moisture dynamics under contrasting
663 vegetation types. *Journal of Hydrology in review*

664 Dolman AJ, Miralles DG, de Jeu RAM. 2014. Fifty years since Monteith’s 1965
665 seminal paper: The emergence of global ecohydrology. *Ecohydrology* **7** (3):
666 897–902 DOI: 10.1002/eco.1505

667 Dore MHI. 2005. Climate change and changes in global precipitation patterns: What
668 do we know? *Environment International* **31** (8): 1167–1181 DOI:
669 10.1016/j.envint.2005.03.004

670 Dunn SM, Mackay R. 1995. Spatial variation in evapotranspiration and the influence
671 of land use on catchment hydrology. *Journal of Hydrology* **171** (1–2): 49–73
672 DOI: 10.1016/0022-1694(95)02733-6

673 Ebel BA. 2013. Simulated unsaturated flow processes after wildfire and interactions
674 with slope aspect. *Water Resources Research* **49** (12): 8090–8107 DOI:
675 10.1002/2013WR014129

676 Engel V, Jobbágy EG, Stieglitz M, Williams M, Jackson RB. 2005. Hydrological
677 consequences of Eucalyptus afforestation in the Argentine Pampas. *Water
678 Resources Research* **41** (10) DOI: 10.1029/2004WR003761

679 Fabris L, Malcolm IA, Buddendorf WB, Millidine KJ, Tetzlaff D, Soulsby C. 2017.
680 Hydraulic modelling of the spatial and temporal variability in Atlantic salmon
681 parr habitat availability in an upland stream. *Science of the Total Environment*
682 **601–602**: 1046–1059 DOI: 10.1016/j.scitotenv.2017.05.112

683 Favreau G, Cappelaere B, Massuel S, Leblanc M, Boucher M, Boulain N, Leduc C.
684 2009. Land clearing, climate variability, and water resources increase in semiarid
685 southwest Niger: A review. *Water Resources Research* **45** (7) DOI:
686 10.1029/2007WR006785

687 Feddes RA, Bresler E, Neuman SP. 1974. Field test of a modified numerical model
688 for water uptake by root systems. *Water Resources Research* **10** (6): 1199–1206
689 DOI: 10.1029/WR010I006P01199

690 Ferguson IM, Jefferson JL, Maxwell RM, Kollet SJ. 2016. Effects of root water
691 uptake formulation on simulated water and energy budgets at local and basin
692 scales. *Environmental Earth Sciences* **75** (4): 316 DOI: 10.1007/s12665-015-
693 5041-z

694 van Genuchten MT. 1980. A Closed-form Equation for Predicting the Hydraulic
695 Conductivity of Unsaturated Soils. *Soil Science Society of America Journal* **44**
696 (5): 892 DOI: 10.2136/sssaj1980.03615995004400050002x

697 van Genuchten MT. 1987. A numerical model for water and solute movement in and
698 below the root zone. Riverside, California.

699 Geris J, Tetzlaff D, McDonnell J, Anderson J, Paton G, Soulsby C. 2015a.
700 Ecohydrological separation in wet, low energy northern environments? A
701 preliminary assessment using different soil water extraction techniques.
702 *Hydrological Processes* **29** (25): 5139–5152 DOI: 10.1002/hyp.10603

703 Geris J, Tetzlaff D, McDonnell J, Soulsby C. 2015b. The relative role of soil type and
704 tree cover on water storage and transmission in northern headwater catchments.
705 *Hydrological Processes* **29** (7): 1844–1860 DOI: 10.1002/hyp.10289

706 Geris J, Tetzlaff D, Soulsby C. 2015c. Resistance and resilience to droughts:
707 Hydropedological controls on catchment storage and run-off response.
708 *Hydrological Processes* **29** (21): 4579–4593 DOI: 10.1002/hyp.10480

709 Gosling R. 2012. Assessing the impact of projected climate change on drought
710 vulnerability in Scotland. *Hydrology Research* **45** (6): 806–817 DOI:
711 10.2166/nh.2014.148

712 Gribb MM, Forkutsa I, Hansen A, Chandler DG, McNamara JP. 2009. The Effect of
713 Various Soil Hydraulic Property Estimates on Soil Moisture Simulations. *Vadose*
714 *Zone Journal* **8** (2): 321 DOI: 10.2136/vzj2008.0088

715 Groisman PY, Karl TR, Easterling DR, Knight RW, Jamason PF, Hennessy KJ,
716 Suppiah R, Page CM, Wibig J, Fortuniak K, et al. 1999. Changes in the
717 Probability of Heavy Precipitation: Important Indicators of Climatic Change.
718 *Climatic Change* **42** (1): 243–283 DOI: 10.1023/A:1005432803188

719 Guan H, Hutson J, Ding Z, Love A, Simmons CT, Deng Z. 2013. Principal
720 component analysis of watershed hydrochemical response to forest clearance and
721 its usefulness for chloride mass balance applications. *Water Resources Research*
722 **49** (7): 4362–4378 DOI: 10.1002/wrcr.20357

723 Haria AH, Price DJ. 2000. Evaporation from Scots pine (*Pinus sylvestris*) following
724 natural re-colonisation of the Cairngorm mountains, Scotland. *Hydrology and*
725 *Earth System Sciences* **4** (3): 451–461 DOI: 10.5194/hess-4-451-2000

726 House AR, Thompson JR, Acreman MC. 2016. Projecting impacts of climate change
727 on hydrological conditions and biotic responses in a chalk valley riparian
728 wetland. *Journal of Hydrology* **534**: 178–192 DOI:
729 10.1016/j.jhydrol.2016.01.004

730 Hrachowitz M, Soulsby C, Imholt C, Malcolm IA, Tetzlaff D. 2010. Thermal regimes
731 in a large upland salmon river: A simple model to identify the influence of
732 landscape controls and climate change on maximum temperatures. *Hydrological*
733 *Processes* **24** (23): 3374–3391 DOI: 10.1002/hyp.7756

734 Huang J, Zhou Y, Hou R, Wenninger J. 2015. Simulation of water use dynamics by
735 *Salix* bush in a semiarid shallow groundwater area of the Chinese Erdos Plateau.
736 *Water (Switzerland)* **7** (12): 6999–7021 DOI: 10.3390/w7126671

737 van Huijgevoort MHJ, Tetzlaff D, Sutanudjaja EH, Soulsby C. 2016. Using high
738 resolution tracer data to constrain water storage, flux and age estimates in a
739 spatially distributed rainfall-runoff model. *Hydrological Processes* **30** (25):
740 4761–4778 DOI: 10.1002/hyp.10902

741 IPCC. 2014. Summary for Policymakers. In *Climate Change 2014: Impacts,*
742 *Adaptation, and Vulnerability. Part A: Global and Sectoral Aspects.*

743 *Contribution of Working Group II to the Fifth Assessment Report of the*
744 *Intergovernmental Panel on Climate Change* 1–32. DOI:
745 10.1016/j.renene.2009.11.012

746 Ivits E, Horion S, Fensholt R, Cherlet M. 2014. Drought footprint on European
747 ecosystems between 1999 and 2010 assessed by remotely sensed vegetation
748 phenology and productivity. *Global Change Biology* **20** (2): 581–593 DOI:
749 10.1111/gcb.12393

750 Jacques D, Šimůnek J, Timmerman A, Feyen J. 2002. Calibration of Richards' and
751 convection-dispersion equations to field-scale water flow and solute transport
752 under rainfall conditions. *Journal of Hydrology* **259** (1–4): 15–31 DOI:
753 10.1016/S0022-1694(01)00591-1

754 Jasechko S, Sharp ZD, Gibson JJ, Birks SJ, Yi Y, Fawcett PJ. 2013. Terrestrial water
755 fluxes dominated by transpiration. *Nature* **496** (7445): 347–350 DOI:
756 10.1038/nature11983

757 Kelly AE, Goulden ML. 2008. Rapid shifts in plant distribution with recent climate
758 change. *Proceedings of the National Academy of Sciences of the United States of*
759 *America* **105** (33): 11823–6 DOI: 10.1073/pnas.0802891105

760 Knighton JO, DeGaetano A, Walter MT, Knighton JO, DeGaetano A, Walter MT.
761 2017. Hydrologic State Influence on Riverine Flood Discharge for a Small
762 Temperate Watershed (Fall Creek, United States): Negative Feedbacks on the
763 Effects of Climate Change. *Journal of Hydrometeorology* **18** (2): 431–449 DOI:
764 10.1175/JHM-D-16-0164.1

765 Koch J, Cornelissen T, Fang Z, Bogena H, Diekkrüger B, Kollet S, Stisen S. 2016.
766 Inter-comparison of three distributed hydrological models with respect to
767 seasonal variability of soil moisture patterns at a small forested catchment.
768 *Journal of Hydrology* **533**: 234–249 DOI: 10.1016/j.jhydrol.2015.12.002

769 Kroes JG, Van Dam JC, Groenendijk P, Hendriks RFA, Jacobs CMJ. 2008. *SWAP*
770 *version 3.2. Theory description and user manual*. Wageningen, Alterra. DOI:
771 ISSN 1566-7197

772 Legesse D, Vallet-Coulomb C, Gasse F. 2003. Hydrological response of a catchment
773 to climate and land use changes in Tropical Africa: Case study south central
774 Ethiopia. *Journal of Hydrology* **275** (1–2): 67–85 DOI: 10.1016/S0022-
775 1694(03)00019-2

776 LeMone M a., Chen F, Alfieri JG, Tewari M, Geerts B, Miao Q, Grossman RL,
777 Coulter RL. 2007. Influence of Land Cover and Soil Moisture on the Horizontal
778 Distribution of Sensible and Latent Heat Fluxes in Southeast Kansas during
779 IHOP_2002 and CASES-97. *Journal of Hydrometeorology* **8** (1): 68–87 DOI:
780 10.1175/JHM554.1

781 Li D, Shao M. 2014. Temporal stability analysis for estimating spatial mean soil water
782 storage and deep percolation in irrigated maize crops. *Agricultural Water*
783 *Management* **144**: 140–149 DOI: 10.1016/j.agwat.2014.05.012

784 Li H, Yi J, Zhang J, Zhao Y, Si B, Hill RL, Cui L, Liu X. 2015. Modeling of soil
785 water and salt dynamics and its effects on root water uptake in Heihe arid
786 wetland, gansu, China. *Water (Switzerland)* **7** (5): 2382–2401 DOI:
787 10.3390/w7052382

788 Li Q, Sun Y, Yuan W, Lyu S, Wan F. 2017. Streamflow responses to climate change
789 and LUCC in a semi-arid watershed of Chinese Loess Plateau. *J Arid Land* **9** (4):
790 609–621 DOI: 10.1007/s40333-017-0095-2

- 791 Van Der Linden S, Virtanen T, Oberman N, Kuhry P. 2003. Sensitivity analysis of
792 discharge in the Arctic Usa basin, East-European Russia. *Climatic Change* **57**:
793 139–161 Available at:
794 <https://link.springer.com/content/pdf/10.1023%2FA%3A1022194026904.pdf>
795 [Accessed 19 September 2017]
- 796 Liu Y, Zhang X, Xia D, You J, Rong Y, Bakir M. 2013. Impacts of Land-Use and
797 Climate Changes on Hydrologic Processes in the Qingyi River Watershed, China.
798 *Journal of Hydrologic Engineering* **18** (11): 1495–1512 DOI:
799 10.1061/(ASCE)HE.1943-5584.0000485
- 800 Love D, Uhlenbrook S, Corzo-Perez G, Twomlow S, van der Zaag P. 2010. Rainfall–
801 interception–evaporation–runoff relationships in a semi-arid catchment, northern
802 Limpopo basin, Zimbabwe. *Hydrological Sciences Journal* **55** (5): 687–703 DOI:
803 10.1080/02626667.2010.494010
- 804 McNamara JP, Chandler D, Seyfried M, Achet S. 2005. Soil moisture states, lateral
805 flow, and streamflow generation in a semi-arid, snowmelt-driven catchment.
806 *Hydrological Processes* **19** (20): 4023–4038 DOI: 10.1002/hyp.5869
- 807 Mekis E, Hogg WD. 1999. Rehabilitation and analysis of Canadian daily precipitation
808 time series. *Atmosphere-Ocean* **37** (1): 53–85 DOI:
809 10.1080/07055900.1999.9649621
- 810 Meng F, Su F, Yang D, Tong K, Hao Z. 2016. Impacts of recent climate change on
811 the hydrology in the source region of the Yellow River basin. *Journal of*
812 *Hydrology: Regional Studies* **6**: 66–81 DOI: 10.1016/j.ejrh.2016.03.003
- 813 Miralles DG, De Jeu RAM, Gash JH, Holmes TRH, Dolman AJ. 2011. Magnitude
814 and variability of land evaporation and its components at the global scale.
815 *Hydrology and Earth System Sciences* **15** (3): 967–981 DOI: 10.5194/hess-15-
816 967-2011
- 817 Nunes AN, de Almeida AC, Coelho COA. 2011. Impacts of land use and cover type
818 on runoff and soil erosion in a marginal area of Portugal. *Applied Geography* **31**
819 (2): 687–699 DOI: 10.1016/j.apgeog.2010.12.006
- 820 Ray D, Morison J, Broadmeadow M. 2010. Climate change impacts and adaptation in
821 England’s woodlands. *FC Research Note* **201**: 1–16 Available at:
822 [https://www.forestry.gov.uk/pdf/FCRN201.pdf/\\$FILE/FCRN201.pdf](https://www.forestry.gov.uk/pdf/FCRN201.pdf/$FILE/FCRN201.pdf) [Accessed
823 7 September 2017]
- 824 Richardson AD, Keenan TF, Migliavacca M, Ryu Y, Sonnentag O, Toomey M. 2013.
825 Climate change, phenology, and phenological control of vegetation feedbacks to
826 the climate system. *Agricultural and Forest Meteorology* **169**: 156–173 DOI:
827 10.1016/j.agrformet.2012.09.012
- 828 Rodriguez-Iturbe I, Porporato A, Laio F, Ridolfi L. 2001. Intensive or extensive use
829 of soil moisture: Plant strategies to cope with stochastic water availability.
830 *Geophysical Research Letters* **28** (23): 4495–4497 DOI: 10.1029/2001GL012905
- 831 Romano N. 2014. Soil moisture at local scale: Measurements and simulations.
832 *Journal of Hydrology* **516**: 6–20 DOI: 10.1016/j.jhydrol.2014.01.026
- 833 Scanlon BR, Reedy RC, Stonestrom DA, Prudic DE, Dennehy KF. 2005. Impact of
834 land use and land cover change on groundwater recharge and quality in the
835 southwestern US. *Global Change Biology* **11** (10): 1577–1593 DOI:
836 10.1111/j.1365-2486.2005.01026.x
- 837 Schaap MG, Leij FJ, Van Genuchten MT. 2001. Rosetta: A computer program for
838 estimating soil hydraulic parameters with hierarchical pedotransfer functions.

839 *Journal of Hydrology* **251** (3–4): 163–176 DOI: 10.1016/S0022-1694(01)00466-
840 8

841 Schilling KE, Chan K-S, Liu H, Zhang Y-K. 2010. Quantifying the effect of land use
842 land cover change on increasing discharge in the Upper Mississippi River.
843 *Journal of Hydrology* **387** (3–4): 343–345 DOI: 10.1016/j.jhydrol.2010.04.019

844 Schlesinger WH, Jasechko S. 2014. Transpiration in the global water cycle.
845 *Agricultural and Forest Meteorology* **189–190**: 115–117 DOI:
846 10.1016/j.agrformet.2014.01.011

847 Schneider S, Jacques D, Mallants D. 2013. Inverse modelling with a genetic algorithm
848 to derive hydraulic properties of a multi-layered forest soil. *Soil Research* **51** (5):
849 372–389 DOI: 10.1071/SR13144

850 Séguis L, Cappelaere B, Milési G, Peugeot C, Massuel S, Favreau G. 2004. Simulated
851 impacts of climate change and land-clearing on runoff from a small Sahelian
852 catchment. *Hydrological Processes* **18** (17): 3401–3413 DOI: 10.1002/hyp.1503

853 Serreze MC, Walsh JE, Chapin FSI, Osterkamp T, Dyurgerov M, Romanovsky V,
854 Oechel WC, Morison J, Zhang T, Barry RG. 2000. Observational evidence of
855 recent change in the northern high- latitude environment. *Climatic Change* **46**
856 (1–2): 159–207 DOI: 10.1023/A:1005504031923

857 Šimůnek J, Hopmans JW. 2002. Parameter Optimization and Nonlinear Fitting. In
858 *Methods of Soil Analysis, Part 1, Physical Methods*, Eds. J. H. Dane and G. C.
859 Topp (ed.). Soil Science Society of America: Madison, WI; 139–157. DOI:
860 10.2136/sssabookser5.4.c7

861 Šimůnek J, van Genuchten MT, Šejna M. 2016. Recent Developments and
862 Applications of the HYDRUS Computer Software Packages. *Vadose Zone*
863 *Journal* **15** (7): 25 DOI: 10.2136/vzj2016.04.0033

864 Šimůnek J, Jacques D, Langergraber G, Bradford SA, Šejna M, Van Genuchten MT.
865 2013a. Numerical modeling of contaminant transport using HYDRUS and its
866 specialized modules. *Journal of the Indian Institute of Science* **93** (2): 265–284
867 DOI: <http://journal.iisc.ernet.in/index.php/iisc/article/view/1224>

868 Šimůnek J, Šejna M, Saito H, Sakai M, van Genuchten MT. 2013b. *The HYDRUS-1D*
869 *Software Package for Simulating the One-Dimensional Movement of Water,*
870 *Heat, and Multiple Solutes in Variably-Saturated Media*. Department of
871 Environmental Sciences University of California Riverside Riverside, California.

872 Soulsby C, Birkel C, Geris J, Dick J, Tunaley C, Tetzlaff D. 2015. Stream water age
873 distributions controlled by storage dynamics and nonlinear hydrologic
874 connectivity: Modeling with high-resolution isotope data. *Water Resources*
875 *Research* **51** (9): 7759–7776 DOI: 10.1002/2015WR017888

876 Soulsby C, Birkel C, Tetzlaff D. 2016a. Modelling storage-driven connectivity
877 between landscapes and riverscapes: towards a simple framework for long-term
878 ecohydrological assessment. *Hydrological Processes* **30** (14): 2482–2497 DOI:
879 10.1002/hyp.10862

880 Soulsby C, Bradford J, Dick J, P. McNamara J, Geris J, Lessels J, Blumstock M,
881 Tetzlaff D. 2016b. Using geophysical surveys to test tracer-based storage
882 estimates in headwater catchments. *Hydrological Processes* DOI:
883 10.1002/hyp.10889

884 Soulsby C, Braun H, Sprenger M, Weiler M, Tetzlaff D. 2017a. Influence of forest
885 and shrub canopies on precipitation partitioning and isotopic signatures.
886 *Hydrological Processes* DOI: 10.1002/hyp.11351

- 887 Soulsby C, Dick J, Scheliga B, Tetzlaff D. 2017b. Taming the flood—How far can we
888 go with trees? *Hydrological Processes* **31** (17): 3122–3126 DOI:
889 10.1002/hyp.11226
- 890 Soulsby C, Tetzlaff D, van den Bedem N, Malcolm IA, Bacon PJ, Youngson AF.
891 2007. Inferring groundwater influences on surface water in montane catchments
892 from hydrochemical surveys of springs and streamwaters. *Journal of Hydrology*
893 **333** (2–4): 199–213 DOI: 10.1016/j.jhydrol.2006.08.016
- 894 Sprenger M, Tetzlaff D, Soulsby C. 2017. Stable isotopes reveal evaporation
895 dynamics at the soil-plant-atmosphere interface of the critical zone. *Hydrology*
896 *and Earth System Sciences Discussions* (February): 1–37 DOI: 10.5194/hess-
897 2017-87
- 898 Sterling SM, Ducharne A, Polcher J. 2012. The impact of global land-cover change
899 on the terrestrial water cycle. *Nature Climate Change* **3** (4): 385–390 DOI:
900 10.1038/nclimate1690
- 901 Steven HM, Carlisle A. 1959. *The Native Pinewoods of Scotland*. Oliver & Boyd:
902 Edinburgh & London.
- 903 Sutanto SJ, Wenninger J, Coenders-Gerrits AMJ, Uhlenbrook S. 2012. Partitioning of
904 evaporation into transpiration, soil evaporation and interception: A comparison
905 between isotope measurements and a HYDRUS-1D model. *Hydrology and Earth*
906 *System Sciences* **16** (8): 2605–2616 DOI: 10.5194/hess-16-2605-2012
- 907 Tetzlaff D, Birkel C, Dick J, Geris J, Soulsby C. 2014. Storage dynamics in
908 hydrogeological units control hillslope connectivity, runoff generation, and the
909 evolution of catchment transit time distributions. *Water Resources Research* **50**
910 (2): 969–985 DOI: 10.1002/2013WR014147
- 911 Tetzlaff D, Buttle J, Carey SK, van Huijgevoort MHJ, Laudon H, McNamara JP,
912 Mitchell CPJ, Spence C, Gabor RS, Soulsby C. 2015. A preliminary assessment
913 of water partitioning and ecohydrological coupling in northern headwaters using
914 stable isotopes and conceptual runoff models. *Hydrological Processes*: n/a-n/a
915 DOI: 10.1002/hyp.10515
- 916 Tetzlaff D, Soulsby C, Buttle J, Capell R, Carey SK, Laudon H, McDonnell J,
917 McGuire K, Seibert J, Shanley J. 2013. Catchments on the cusp? Structural and
918 functional change in northern ecohydrology. *Hydrological Processes* **27** (5):
919 766–774 DOI: 10.1002/hyp.9700
- 920 Tetzlaff D, Soulsby C, Waldron S, Malcolm IA, Bacon PJ, Dunn SM, Lilly A,
921 Youngson AF. 2007. Conceptualization of runoff processes using a geographical
922 information system and tracers in a nested mesoscale catchment. *Hydrological*
923 *Processes* **21** (10): 1289–1307 DOI: 10.1002/hyp.6309
- 924 Tudor, GJ, Shewry, MC, Mackey, Ec, Elston, Da, Underwood, FM. 1999. Land cover
925 change in Scotland: the methodology of the Natural Countryside Monitoring
926 Scheme Available at: <http://www.snh.org.uk/pdfs/publications/research/127.pdf>
927 [Accessed 27 November 2017]
- 928 Verstraeten WW, Muys B, Feyen J, Veroustraete F, Minnaert M, Meiresonne L, De
929 Schrijver A. 2005. Comparative analysis of the actual evapotranspiration of
930 Flemish forest and cropland, using the soil water balance model WAVE.
931 *Hydrology and Earth System Sciences* **9**: 225–241 DOI: 10.5194/hess-9-225-
932 2005
- 933 Viviroli D, Weingartner R, Messerli B. 2003. Assessing the Hydrological
934 Significance of the World’s Mountains. *Mountain Research and Development* **23**
935 (4): 369–375 DOI: 10.1659/0276-4741(2003)023

936 Vivoni ER, Rinehart AJ, Méndez-Barroso LA, Aragón CA, Bisht G, Cardenas MB,
937 Engle E, Forman BA, Frisbee MD, Gutiérrez-Jurado HA, et al. 2008. Vegetation
938 controls on soil moisture distribution in the Valles Caldera, New Mexico, during
939 the North American monsoon. *Ecohydrology* **1** (3): 225–238 DOI:
940 10.1002/eco.11

941 Walther G-R. 2010. Community and ecosystem responses to recent climate change.
942 *Philosophical transactions of the Royal Society of London. Series B, Biological*
943 *sciences* **365** (1549): 2019–2024 DOI: 10.1098/rstb.2010.0021

944 Wang H, Guan H, Guyot A, Simmons CT, Lockington DA. 2016a. Quantifying
945 sapwood width for three Australian native species using electrical resistivity
946 tomography. *Ecohydrology* **9** (1): 83–92 DOI: 10.1002/eco.1612

947 Wang H, Guan H, Simmons CT. 2016b. Modeling the environmental controls on tree
948 water use at different temporal scales. *Agricultural and Forest Meteorology* **225**:
949 24–35 DOI: 10.1016/j.agrformet.2016.04.016

950 Wang H, Tetzlaff D, Dick JJ, Soulsby C. 2017a. Assessing the environmental controls
951 on Scots pine transpiration and the implications for water partitioning in a boreal
952 headwater catchment. *Agricultural and Forest Meteorology* **240–241**: 58–66
953 DOI: 10.1016/j.agrformet.2017.04.002

954 Wang H, Tetzlaff D, Soulsby C. 2017b. Testing the Maximum Entropy Production
955 approach for estimating evapotranspiration from closed canopy shrubland in a
956 low-energy humid environment. *Hydrological Processes* DOI:
957 10.1002/hyp.11363

958 Wang J, Bras RL. 2011. A model of evapotranspiration based on the theory of
959 maximum entropy production. *Water Resources Research* **47** (3): n/a-n/a DOI:
960 10.1029/2010WR009392

961 Xu YP, Zhang X, Ran Q, Tian Y. 2013. Impact of climate change on hydrology of
962 upper reaches of Qiantang River Basin, East China. *Journal of Hydrology* **483**:
963 51–60 DOI: 10.1016/j.jhydrol.2013.01.004

964 Yang Y, Guan H, Shen M, Liang W, Jiang L. 2015. Changes in autumn vegetation
965 dormancy onset date and the climate controls across temperate ecosystems in
966 China from 1982 to 2010. *Global change biology* **21** (2): 652–65 DOI:
967 10.1111/gcb.12778

968 Zhang J, Felzer BS, Troy TJ. 2016. Extreme precipitation drives groundwater
969 recharge: the Northern High Plains Aquifer, central United States, 1950-2010.
970 *Hydrological Processes* **30** (14): 2533–2545 DOI: 10.1002/hyp.10809

971 Zhang L, Dawes WR, Walker GR. 2001. Response of mean annual evapotranspiration
972 to vegetation changes at catchment scale. *Water Resources Research* **37** (3):
973 701–708 DOI: 10.1029/2000WR900325

974 Zhang L, Hu Z, Fan J, Zhou D, Tang F. 2014. A meta-analysis of the canopy light
975 extinction coefficient in terrestrial ecosystems. *Frontiers of Earth Science* **8** (4):
976 599–609 DOI: 10.1007/s11707-014-0446-7
977