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Dynamic contributions of stratified groundwater to streams controls seasonal variations of streamwater transit times

Jean Marçais^{1,2}, Louis A. Derry^{3,2}, Luca Guillaumot^{4,5}, Luc Aquilina⁴,
Jean-Raynald de Dreuzy^{4,6}

¹INRAE, UR Riverly, F-69625 Villeurbanne, France

²Institut de Physique du Globe de Paris, Université de Paris, CNRS, F-75005 Paris, France

³Department of Earth and Atmospheric Sciences, Cornell University, Ithaca, NY

⁴Univ Rennes, CNRS, Géosciences Rennes - UMR 6118, F-35000 Rennes, France

⁵International Institute for Applied Systems Analysis, Laxenburg, Austria

⁶Univ Rennes, CNRS, OSUR Observatoire des sciences de l'univers de Rennes - UMS 3343, F-35000
Rennes, France

Key Points:

- Hydraulic conductivity controls groundwater flow contributions to streams while porosity directly scales the streamwater mean transit times.
- Seasonal groundwater contributions to streams modulate the mean transit times dynamics from 6 years at high flows to 20 years at low flows.
- Stratified groundwater ages lead to a significant fraction (> 75 %) of old water to streamflow (>1 year).

Corresponding author: Jean Marçais, jean.marcais@inrae.fr

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Abstract

Streamwater transit time distributions display a variable proportion of old waters (≥ 1 yr). We hypothesize that the corresponding long transit times result from groundwater contributions to the stream and that seasonal streamwater transit time variations result from 1) the variable contributions of different flowpaths (overland flow, seepage flow and baseflow) and 2) the stratification of groundwater residence times. We develop a parsimonious model to capture the groundwater contribution to the stream discharge and its effect on transient transit times. Infiltration is partitioned according to the aquifer saturation between Boussinesq groundwater flow and overland flow. Time-variable transit time distributions are obtained with a new 2D particle tracking algorithm. Hydraulic conductivity, total and drainable porosities are calibrated by using discharge and CFC tracer data on a crystalline catchment located in Brittany (France). The calibrated models succeed in reproducing CFCs concentrations and discharge dynamics. The groundwater flow contribution to the stream is controlled by the aquifer hydraulic conductivity, while its age is controlled by the drainable and total porosities. Old groundwater (≥ 1 yr) is the source for approximately 75 % of the streamflow with strong seasonal variations (between 40 and 95 %). Mean transit times are approximately 13 years, varying between 6 and 20 years, inversely proportional to the groundwater contribution. These seasonal variations are driven by the groundwater versus overland flow partitioning. The stratification of groundwater residence times in the aquifer plays a minor role in the streamwater transit times but is key for the transit time dynamics of the groundwater contribution to the stream.

Plain Language Summary

Water entering a catchment as precipitation can take multiple paths with different transit times to the stream. While a significant fraction of water has short transit times (i.e. is “young”) when it reaches the stream, there is also an important contribution of “old” water with long transit times in the subsurface. The age distribution of this old component is important for understanding the resilience of watersheds to climate change, the behavior of persistent pollutants and chemical weathering processes. We developed a model, informed with discharge time series and atmospheric age tracer (CFCs), to constrain the age distribution in both groundwater and streamwater and how they vary seasonally. In our temperate test catchment with crystalline bedrock (Brittany, France), we find that the mean age of streamwater exiting the catchment is approximately 13 years, but varies seasonally from 6 to 20 years as the relative contributions of older groundwater and younger runoff change. Groundwater stratification further influences the mean age of the groundwater contribution arriving at the stream.

1 Introduction

Transit time is the time elapsed between the moment a water molecule enters a hydrological system (e.g. a catchment) and the moment it exits (e.g. through discharge) (Sprenger et al., 2019). This quantity is a fundamental descriptor of the hydrological fluxes and transport within the Critical Zone (Botter et al., 2020). Sampled streamwater, comprised of many water molecules with different transit times, reflects the diversity of flowpaths to the stream and its age distribution can provide information about Critical Zone structures (Kirchner et al., 2000, 2001; Brantley et al., 2017; Kuppel et al., 2020). This

64 distribution of transit times, though not directly observable, is a powerful integral quan-
65 tity for understanding catchment-scale processes (Ameli et al., 2017; Gabrielli et al., 2018;
66 Heidbüchel et al., 2020). It varies at different time scales, from the event to the seasonal
67 and interannual time scales, reflecting the interplay between climatic variations and catch-
68 ment structures as well as their impact on water resources at the catchment scale (Heidbüchel
69 et al., 2013; Remondi et al., 2018, 2019). Characterizing streamwater transit time dis-
70 tributions (TTDs) is therefore critical to define sustainable water use (Ferguson et al.,
71 2020), evaluate climate change impacts on water resources (Maxwell et al., 2016; Fang
72 et al., 2019), assess the effect of contaminant legacy (e.g. agricultural nutrient and pes-
73 ticides) (van der Velde et al., 2010; Van Meter et al., 2016, 2018; Ehrhardt et al., 2019),
74 as well as better understand Critical Zone evolution through weathering and erosion (Maher,
75 2010; Maher & Druhan, 2014; Li et al., 2017; R. S. Anderson et al., 2019).

76 While the distribution of transit times is weighted toward young ages (≤ 1 year)
77 in most catchments (Jasechko et al., 2016; Benettin, Bailey, et al., 2017; Harman, 2014),
78 it also includes a variable and difficult-to-constrain contribution from older water (≥ 1
79 year) investigated by only few studies (Stolp et al., 2010; Kaandorp et al., 2018; Visser
80 et al., 2019). This old water fraction that has experienced a long transit time in the hy-
81 drologic system may be a smaller component of stream fluxes (Gardner et al., 2011), but
82 is critical to constrain for several reasons. These include but are not limited to the fol-
83 lowing. 1) The resilience of surface water fluxes in the face of climate variability (drought,
84 climate change) can depend heavily on the availability of old groundwater contributions
85 to streamflow (Manning et al., 2012; Solder et al., 2016). 2) The recovery time scale of
86 a watershed with persistent chemical pollutants can be alleviated by the old water con-
87 tribution due to dilution or reactive processes (Aquilina et al., 2012; Pinay et al., 2015;
88 Van Meter & Basu, 2017; Kolbe et al., 2019). 3) Chemical weathering (water-rock or water-
89 regolith) reactions are often kinetically limited and water components with long reac-
90 tion time scales are necessary to produce quasi-equilibrium solute and tracer composi-
91 tions (Maher, 2011; Fernandez et al., 2021). Thus “older” water may be a smaller con-
92 tributor to stream discharge but can have a disproportionate influence on weathering re-
93 actions and streamwater quality (Frisbee et al., 2013). Predicting the future response
94 of watersheds to anthropogenic climate forcing requires that we have an improved abil-
95 ity to accurately model the old component of the distribution of watershed transit times
96 and how that distribution varies in response to hydrologic and landscape forcing (Heidbüchel
97 et al., 2012; Wilusz et al., 2017).

98 Here we focus on characterizing old streamwater ages (≥ 1 year). We hypothesize
99 that :

- 100 • old streamwater transit times result from the high contributions of subsurface aquifers
101 and from strong groundwater flow residence time stratification (i.e. young ages
102 at the top of the aquifer and old ages at the bottom, see Figure 1), tied to the hy-
103 draulic properties of the aquifer (porosity, hydraulic conductivity);
- 104 • the dynamic partitioning of discharge between groundwater flows and overland flows
105 explains the high seasonal variability of river discharge while the old component
106 of the streamwater does not evolve much.

107 To assess the relevance of these two processes (i.e. stratification and partitioning)
108 for old mean streamwater transit times and their seasonal variability, we could not use

109 previously developed families of transit time models. The first family of models spans
 110 all the lumped representation of transit time distribution, from the simple lumped pa-
 111 rameters models (Małozzewski & Zuber, 1982; Marçais et al., 2015) to more advanced
 112 storage selection function approaches (Harman, 2014; Rinaldo et al., 2015; Benettin, Soulsby,
 113 et al., 2017; Benettin & Bertuzzo, 2018; Nguyen et al., 2021). This type of models are
 114 not appropriate here because they do not account for spatially explicit processes, which
 115 limit their ability to investigate the controls regarding transit time distributions shapes
 116 (Gauvain et al., 2021), as well as the impact of different flowpaths contributions to streams
 117 on the development of transit time distributions. The second family of models encom-
 118 passes detailed, spatially explicit process-based models such as Fatichi et al. (2016); X. Yang
 119 et al. (2018); Therrien and Sudicky (1996); Kollet and Maxwell (2006); Camporese et
 120 al. (2010). Their explicit description requires constraints on a number of parameters that
 121 can be difficult to obtain in many circumstances, and whose calibration may not be unique.
 122 Besides, due to their high computation requirements, systematic exploration of the vari-
 123 ability of response of these models depending on different parameters sets and on main
 124 processes represented are often not carried out (Putti & Paniconi, 2004; Paniconi & Putti,
 125 2015). We therefore develop a novel intermediate model that represents what we hypoth-
 126 esize are the key processes playing a first order role for characterizing old streamwater
 127 ages and the proportion of old groundwater in streams.

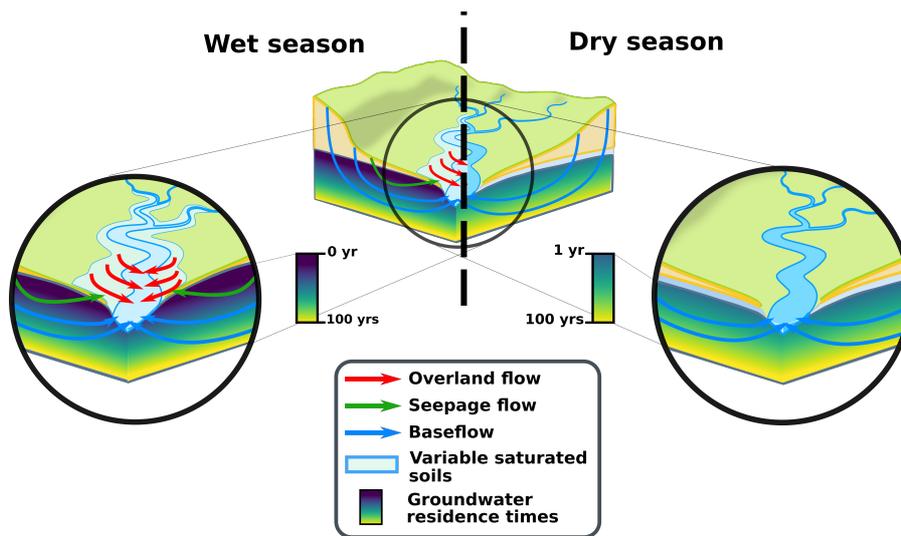


Figure 1. Sketch of the investigated processes. Where the water table intersects the land surface, it generates seepage flow (green arrows) and creates variable saturated areas that are responsive to precipitation (overland flow, red arrows). In the aquifer, groundwater flow is modeled with a Boussinesq formulation (blue arrows) adapted to take into account the groundwater residence time stratification (materialized by the colormap in the aquifer). Note the reduction of overland and seepage flows (red and green arrows) as well as the flush of young groundwater residence times from the wet (left side of the panel) to the dry season (right side of the panel) due to the small displacement of the water table.

128 We use a semi-explicit hillslope-scale approach that represents groundwater flows
 129 inputs to streamwater with the vertically integrated approach proposed by Boussinesq
 130 (Boussinesq, 1877; Dupuit, 1863). We modified the Boussinesq model to take into ac-

count the dynamic partitioning with the saturation excess overland flow and its fast routing to the stream network (S. P. Anderson et al., 1997; Freer et al., 2002; Blume & van Meerveld, 2015). We assume that the partitioning between groundwater flow and overland flow is controlled by the existence of seepage areas when the water table intersects the land surface (section 2.1, Marçais et al. (2017)). To represent the effect of the groundwater residence time stratification (Vogel, 1967; Kolbe et al., 2020), we develop a new transient and stratified transport model (section 2.2) adapted to the hillslope Boussinesq equations (Troch et al., 2003). Further, to overcome classic tracer resolution bias towards the young ages (Kirchner, 2016; Sprenger et al., 2019), we used CFC groundwater age data conjointly with monthly discharge time series to both quantify the groundwater contribution to stream and the transit time distribution dynamics (section 2.3). We test this approach on the 43 km² Guillec catchment of Brittany (France) as an example of shallow aquifers in crystalline regions (section 3). We discuss the relevance of such stratification and partitioning processes, the representativity of the calibrated catchment-scale parameters (hydraulic conductivity, drainable and total porosities), and the potential interest and limitations of the approach (section 4).

2 Flow model, transport model and study site

We recall the flux formalism relative to the hillslope storage Boussinesq equation modified to include overland flow generation (Marçais et al., 2017) (Figure 1). Advective transport is represented within a Lagrangian framework and solved with a particle tracking method accounting for the stratification of the solute transport in the aquifer and the routing of the saturation excess overland flow. It allows tracking of particle transit times and travel distances. Dispersive transport is added within the flow lines with an inverse Gaussian distribution. The model computes estimates of time-varying streamflows, transit time distributions of streamwater and residence time distributions of groundwater (Marçais, 2021) and these are compared to flow and transport data of the Guillec catchment (Brittany, France).

2.1 Flow model: partitioning groundwater from overland flow

Groundwater flows are based on a Boussinesq formalism for unconfined shallow aquifers (Troch et al., 2003; Paniconi et al., 2003) modified to account for saturation excess overland flow generation (Marçais et al., 2017). For a hillslope described by the width function ω , the slope of the hillslope bedrock θ , the aquifer thickness d and its hydraulic parameters, the drainable porosity ϕ_d and the hydraulic conductivity k (Figure 2.A), groundwater flows draining to a stream are modeled by:

$$\left\{ \begin{array}{l} \frac{\partial S}{\partial t}(x, t) + q_S(x, t) = -\frac{\partial Q}{\partial x}(x, t) + N(t)\omega(x) \\ q_S(x, t) = \mathcal{G}\left(\frac{S(x, t)}{S_c(x)}\right) \mathcal{R}\left(-\frac{\partial Q}{\partial x}(x, t) + N(t)\omega(x)\right) \\ Q(x, t) = -\frac{kS(x, t)}{\phi_d} \left(\cos\theta \frac{\partial}{\partial x} \left(\frac{S}{\phi_d w} \right)(x, t) + \sin\theta \right) \\ S(x, t) = \phi_d \omega(x) h(x, t) \\ 0 \leq S(x, t) \leq S_c(x) = \phi_d \omega(x) d(x). \end{array} \right. \quad (1)$$

where t [T] is the time and x [L] the distance to the channel varying from 0 at the river to L at the water divide. $S(x, t)$ [L^2] is the groundwater storage per unit length dx varying from 0 to the maximum groundwater storage $S_c(x)$ [L^2], $Q(x, t)$ [L^3/T] is the integrated Boussinesq groundwater flux, $N(t)$ [L/T] is the infiltration time series and $q_S(x, t)$ [L^2/T] is the saturation excess overland flow generated per unit length dx . $\mathcal{G}(u) = \mathcal{H}(u-1)$ and $\mathcal{R}(u) = u \mathcal{H}(u)$ are two functionals depending on the relative saturation (S/S_c) and flow balance ($-\frac{\partial Q}{\partial x}(x, t) + N(t)\omega(x)$) based on the Heaviside step function \mathcal{H} to control the partition of incoming flux between water storage evolution and saturation excess overland flow (Figure 2.B, Marçais et al. (2017)).

The first equation of system 1 is the expression of the mass conservation for hillslope storage Boussinesq models (Troch et al., 2003). The second equation drives the local occurrence of saturation excess overland flow, when (1) groundwater storage saturates the whole aquifer column up to the surface $\mathcal{G}(\frac{S(x,t)}{S_c(x)}) = 1$ (2) positive incoming flux locally occurs, $-\frac{\partial Q}{\partial x}(x, t) + N(t)\omega(x) \geq 0$. At these conditions, the functional \mathcal{R} is equivalent to the identity function (Figure 2.B) and $q_S(x, t) = -\frac{\partial Q}{\partial x}(x, t) + N(t)\omega(x)$. The third equation expresses the integrated groundwater flux $Q(x, t)$ over the transect made up of a Darcy flow driven by the gradient of storage and of a gravity flow function of the bedrock slope. Fourth equation relates the groundwater storage $S(x, t)$ to the hydraulic head $h(x, t)$. The last equation states that the groundwater storage S cannot be greater than S_c , since $h(x, t)$ is limited by the maximum aquifer depth $d(x)$. Runoff occurrence thus appears to be both bottom-up controlled by the saturation of the aquifer, through the development of variably saturated source areas and top-down controlled by the occurrence of infiltration events.

To take into account the routing process of the saturation excess overland flow and seepage flow, from the hillslope surface to the river, we developed a 1D gravity driven routing model (Text S1). In this study, as we focus on monthly discharge and transit time distribution dynamics, we consider that the discharge in the river can be approximated by $Q_{mod}(x, t) = Q(x, t) + \int_0^L q_S(x, t)dx$, assuming that, at the monthly timescale, saturation excess overland flow can be considered to be routed to the stream on much smaller time scales (Musy & Higy, 2004).

2.2 Transport model: stratifying groundwater flowpaths

2.2.1 A 2D stratified, advective transport

2D advective transport is modeled within a Lagrangian framework defining the flow lines over the whole domain. Horizontal groundwater particles velocity $v_x(x, t)$ directly derives from the Darcy flux Q integrated over the groundwater column (equation 3 of system 1), divided by the surface area of the porous section $\phi_{tot}\omega h$, where ϕ_{tot} is the total aquifer porosity. Replacing the aquifer depth h by the groundwater storage S using the penultimate equation of system 1, we get :

$$v_x(x, t) = \frac{\phi_d}{\phi_{tot}} \frac{Q(x, t)}{S(x, t)} \quad (2)$$

By applying mass conservation principles to the hillslope Boussinesq equations, we retrieve the vertical component of the particle velocities v_z and obtain the following La-

207 grangian system of equations (see Appendix A):

$$208 \left\{ \begin{array}{l} S(x(t), t) \frac{dx}{dt}(t) = \frac{\phi_d}{\phi_{tot}} Q(x(t), t) \\ \frac{dz}{dt}(t) = -z(t) \frac{\phi_d}{\phi_{tot}} \left(\frac{d}{dx} \left(\frac{Q(x(t), t)}{S(x(t), t)} \right) + \frac{1}{w} \frac{dw}{dx} \frac{Q(x(t), t)}{S(x(t), t)} \right) \\ x(t_0) = x_0 \\ z(t_0) = z_0 = h(x_0, t_0) \end{array} \right. \quad (3)$$

209 where t_0 is the initial time, $\mathbf{x}_0 = (x_0, z_0)$ is the initial position of the flow line start-
 210 ing at the water table and $\mathbf{x}(t) = (x(t), z(t))$ is the 2D flow-line trajectory in the aquifer.
 211 With this novel description, the vertical component of the velocity field is derived from
 212 the horizontal velocity variations. Vertical transport of solutes is logically induced by
 213 variations of velocity in the other horizontal directions.

214 The transport equation (3) is solved by a particle method where $N_P = N_x N_t$ par-
 215 ticles are injected for each of the N_t time steps at each of the N_x discretized elements.
 216 Each particle is seeded at the top of the saturated zone and is flux-weighted according
 217 to the intensity of the infiltration flux and to the relative area of the mesh cell (Figure
 218 S1). Particles are injected in the aquifer where aquifer saturation remains below the land
 219 surface, and at the surface otherwise.

220 For particles injected in the aquifer, this eventually leads to an aquifer stratifica-
 221 tion where transit times increase with depth. Older particles are deeper in the aquifer
 222 while the youngest ones, which have been just introduced through the most recent in-
 223 filtration events, are above, closer to the water table (Figure 2). At each time t , parti-
 224 cles crossing the groundwater boundaries (i.e. the water table and the imposed head bound-
 225 ary condition at the river) go from the groundwater compartment directly to the seep-
 226 age areas or to the river. They correspond respectively to the seepage flow and to the
 227 baseflow (Figure 2). Their transit time is essentially equal to the time spent in the sub-
 228 surface compartment $\tau_{gw}(t)$, as the transit time at the surface is much smaller. Parti-
 229 cles injected at the surface are then assumed to have a transit time proportional to the
 230 distance to the stream from where they have been injected:

$$231 \tau_{surf}(x, t) = \frac{x}{L_{sat}} \min_{1 \leq i \leq N(t)} \tau_{gw}^i(t) \quad (4)$$

232 where $\tau_{surf}(x, t)$ is the transit time of the particles injected directly at the surface, at
 233 a distance x from the river, $L_{sat}(t)$ is the length of the saturated extension, and $N(t)$
 234 is the number of groundwater particles exiting at time t .

235 Due to the stratification obtained by the 2D trajectory description, seepage flow
 236 is fed first by the youngest particles, which are at the top of the groundwater column.
 237 Compared to other assumptions of full mixing over the water column, this method lo-
 238 cates the streamline vertically, stratifies the transit times and correlates the depth of the
 239 streamline to the distance of its origin. Deeper streamlines originate from farther away.
 240 It has significant consequences on transit times. The youngest particles are the first to
 241 exit at the onset of the seepage area. The oldest particles are the last to leave and exit
 242 directly in the river. Overall the stratification increases the span of the transit time dis-
 243 tribution. Equation (3) is solved in parallel with an implicit ode solver capable of man-
 244 aging the presence of a singularity in the case where $S(\mathbf{x}(t), t) = 0$. This is achieved

245 with the variable time step and variable order ode15s MATLAB[®] solver (Shampine et
 246 al., 1999). The numerical methods are validated by comparing the flux-based modeled
 247 discharge Q_{mod} with the particular modeled discharge obtained by computing the par-
 248 ticle flux reaching the stream at each time step (TextS2 and Figure S2). Results can be
 249 expressed as particle trajectories.

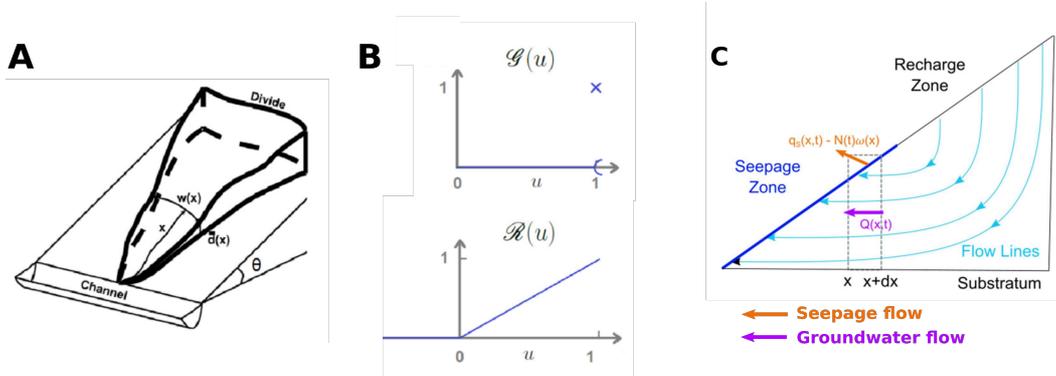


Figure 2. Sketch of A. the hillslope representation and geomorphologic variables ($\omega(x)$, the hillslope width, $d(x)$, the aquifer depth and $\theta(x)$, the hillslope bedrock slope), reproduced from Troch et al. (2003). B. the functions \mathcal{G} and \mathcal{R} used in the partitioning system of equations (1). C. the streamline organization. The stratification of the flow lines derived from the particle tracking method respects the conservation of mass laterally (in the direction modeled by the flux equation 1) and vertically through the premise of "no crossing" between streamlines. The strategy to retrieve seeping particles can be summarized as "last in, first out" for the particles, therefore suggesting a process-based interpretation for the preferential selection of young water discharged to the stream and to the seepage zone.

250 In the Boussinesq approach of groundwater flows, celerity (flux response to precipi-
 251 tation) is dissociated from velocity (effective particle speed inside the aquifer) due to
 252 three effects. First, the celerity response in time is only dictated by $Q(x, t)$ while the ve-
 253 locity response is driven by the ratio $\frac{Q(x, t)}{S(x, t)}$, meaning that the dynamics of stored drain-
 254 able water also affect transport properties (Scaini et al., 2017). Second, the presence of
 255 a non-linearity due to the water table intersection with the land surface creates a respon-
 256 sive zone (the variable saturated areas) for flows, while the transport response is still pre-
 257 dominantly controlled by the transit times of the particles travelling in the groundwa-
 258 ter compartment. Third, it is the total amount of groundwater stored that effectively
 259 conditions the particle velocity by modifying the Darcy velocity with the ratio $\frac{\phi_d}{\phi_{tot}}$ (van
 260 Verseveld et al., 2017). In this model, these three effects contribute to dissociate celer-
 261 ity from velocity (McDonnell & Beven, 2014; McDonnell, 2017).

262 *2.2.2 A longitudinal advective-dispersive transport scheme to model time* 263 *variable residence time and transit time distributions*

264 The numerical scheme introduced in the previous section corresponds to the 2D
 265 purely advective transport of a solute without any diffusion or dispersion. We introduce
 266 longitudinal diffusive and dispersive processes in the aquifer by considering that trans-

port within each flow line can be modeled by an inverse Gaussian distribution accounting both for advective and diffusive/dispersive processes (Engdahl & Maxwell, 2014; Kirchner et al., 2001; de Marsily, 1986) (Text S3 and Figure S3). Dispersive and diffusive processes transversal to the flow lines are neglected because it is considered one order smaller than longitudinal dispersivity (de Marsily, 1986) and of second-order effects in such uniform media (de Dreuzy et al., 2007). For each particle with an age τ (which can be considered as a transit time or as a residence time depending on the medium sampled, stream or catchment), we obtain the following expressions for the distribution of transit times $p(t, \tau)$:

$$\begin{cases} p(t, \tau) = \sqrt{\frac{25\tau}{4\pi t^3}} \exp\left(-\frac{25(t-\tau)^2}{4\tau t}\right) & \text{if } t \geq 0 \\ p(t, \tau) = 0 & \text{if } t < 0 \end{cases} \quad (5)$$

The residence time distribution at time t can be derived from the residence time of all the particles k located in the aquifer and at the surface. The transit time distribution at time t is defined from the transit time of the particles i reaching the river between t and $t+\Delta t$. In both cases, the distribution of times is obtained by summing up the inverse Gaussian distributions centered on their individual traveling time τ^i in the aquifer and at the surface. Each of these inverse Gaussian distribution is weighted by the water volume associated to the particle. This results in synthetic, physically-based transient residence time distributions $p_S(t, \tau)$ and transit time distributions $p_Q(t, \tau)$.

2.3 Guillec catchment: field data and modeling strategy

To illustrate the value of this modeling approach, we assessed its relevance on the Guillec catchment located in western Brittany (France). The Guillec catchment has a relatively limited size (43 km²). Discharge and CFC data are available. The Guillec catchment is characterized by an oceanic temperate climate with average precipitation of around 1000 mm/yr and real evapotranspiration approximately equal to 600 mm/yr. Its geology is crystalline, mainly granitic, though some siliclastic sedimentary materials are present in the fluvial system due to the weathering of the crystalline bedrock. Topographic gradients are relatively limited with altitudes ranging between 7 m and 125 m (Figure 3).

2.3.1 Flux data: monitored discharge and modeled infiltration data

Daily discharge time series are available over 54 years at the catchment outlet. With the objective to analyze the interannual and seasonal fluctuations of discharge and time-varying transit time distributions, the monthly average of this time series is derived. Input forcings for our modeling strategy come from the SURFEX land surface model (Le Moigne et al., 2020) seeded with SAFRAN atmospheric reanalysis (Vidal et al., 2010). SAFRAN provides precipitation and energy atmospheric forcings over France at the unified national scale (8x8 km² grid) (Quintana-Seguí et al., 2008). The SURFEX land surface model computes water and energy budgets in the soil compartment and produces precipitation, real evapotranspiration, runoff (shallow runoff) and drainage (deep infiltration) on the same national grid scale (Boone et al., 2017). To stick to our modeling framework, we summed runoff and drainage time series to provide a proper infiltration time series as the model itself spatially resolves the partitioning between infiltration excess overland flow and recharge (Marçais et al., 2017).

2.3.2 *Transport data: sampling and interpretation of CFCs*

Depending on the timescale of interest, an appropriate age tracer has to be chosen for groundwater dating (Suckow, 2014; Kazemi et al., 2006). Previous modeling studies in Brittany have estimated the mean streamwater transit times to be approximately 5 years to 7 years (Martin et al., 2006; Mougin et al., 2015). Groundwater mean residence times, which have been evaluated from CFC sampling in aquifers, show an older range, between 25 years (Marçais et al., 2018; Roques et al., 2014; Ayraud et al., 2008) and 50 years (Kolbe et al., 2016). Groundwater inputs to streams are significant, from 30 % to 60 % of the stream discharge (Clément et al., 2003; Mougin et al., 2002, 2008; Aquilina et al., 2012). A first-order analysis shows apparent inconsistencies between these observations at least when they are compared to regional scale estimates, as a 50 % input of 25 year old groundwater to streams would lead to streamwater mean transit times of at least 12.5 years. As we are interested in understanding the contribution and age of groundwater to the Guillec river, we chose CFCs to date groundwater because CFC ages are sensitive to time scales of years to decades.

CFCs are assumed to be in equilibrium with the atmosphere in the unsaturated zone, provided that the water table is not too deep (typically not more than 5m)(Engesgaard et al., 2004), which is generally the case for Brittany (Mougin et al., 2004; Kolbe et al., 2016). Once the infiltrating water recharges the aquifer, it becomes isolated from the atmosphere, with the dissolved concentration equilibrated with the atmospheric concentration at the time of isolation. CFCs are then assumed to be transported conservatively in the aquifer. For this analysis, we sampled the two wells shown as violet circles on Figure 3A and the Guillec river close to its outlet (blue triangle). We calibrated the transport model on these three CFCs samples. Aerobic conditions prevent CFC degradation (Dunkle et al., 1993; Kolbe et al., 2020). In any case, exchanges between the samples and the atmosphere have been minimized, by sampling far from the river re-aeration zones and at the very bottom of the water column (Vautier et al., 2020; Guillaumot et al., 2021). The two wells were chosen because they are known to be located just upstream of well-identified springs. Sampled waters are thus confirmed to characterize the contribution of the aquifer to the stream. In the river, we sampled CFCs at the end of the low flow period in October to ensure both significant river depth and a majority of streamwater coming from the aquifer.

CFCs were analyzed at the CONDATE Eau facility of the University of Rennes, France (OSUR laboratory) (Ayraud et al., 2008; Labasque et al., 2014). CFC-11 and CFC-113 concentrations were found to be consistent with each other and were retained as modeling targets, except one contaminated CFC-11 sample (Table S1). In contrast, most CFC-12 concentrations appeared to be contaminated as the values were well above the maximum atmospheric concentration ever recorded.

2.3.3 *Topographic data: an equivalent hillslope representation*

The hillslope model has been derived on the basis of the topographically defined catchment from the gauging station at the outlet (Figure 3A). SRTM data (Farr et al., 2007) were analyzed with the topotoolbox package (Schwanghart & Scherler, 2014) under the assumption that the hydrological and hydrogeological catchments correspond (Reggiani et al., 1998; Reggiani & Schellekens, 2003; Haitjema & Mitchell-Bruker, 2005; Gleeson & Manning, 2008). This assumption is generally appropriate in Brittany where limited

353 weathering of the crystalline bedrocks and low topographical gradients preclude the ex-
 354 istence of significant regional groundwater circulation (Kolbe et al., 2016). We chose to
 355 rely on a representation of the catchment based on an equivalent hillslope representa-
 356 tion characterized by its relative elevation to the river (Loritz et al., 2017; Reggiani &
 357 Schellekens, 2003). This equivalent hillslope is obtained by aggregating the cells of the
 358 DEM in a 1D representation according to their distance to the river in the DEM (Fig-
 359 ure 3B and C). The slope of its surface is relatively limited to be around 2.5 %. The 2D
 360 DEM-based analysis gives the width function and the mean elevation function at the scale
 361 of the catchment. Assuming a flat bedrock located at the same elevation as the river out-
 362 let (L’vovich, 1979), the width and elevation functions characterize the storage capac-
 363 ity of the catchment S_{tot} (which we assumed to be mainly located in its shallow aquifer)
 364 as a function of the distance to the stream: $S_{tot}(x, t) = \phi_{tot} \omega(x) d(x)$.

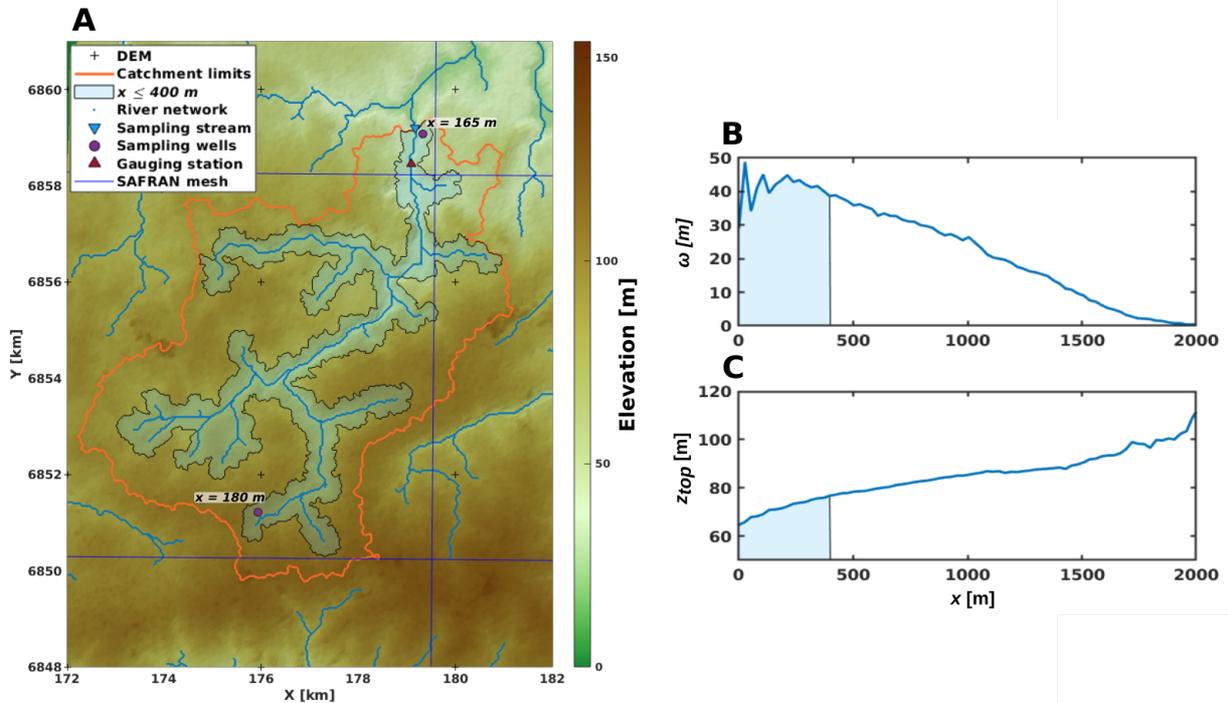


Figure 3. (A) 2D representation of the topography of the Guillec catchment with the sampling locations indicated as circles and triangles. The distance of the sampled locations to the river are indicated in black. The gridding mesh used to retrieve infiltration time series from the SURFEX model is superimposed in blue. (B) Width function $\omega(x)$ and (C) elevation function $z_{top}(x)$ of the equivalent hillslope characterizing the repartition of the catchment area as a function of the distance x to the stream. To illustrate the aggregation strategy, we highlight the elevation and the width of the equivalent hillslope in pale blue for a distance to the stream smaller than 400 m. Corresponding DEM cells are highlighted with the same color in (A).

365 2.3.4 Model calibration

366 The model has 3 parameters to calibrate, which are the bedrock hydraulic conduc-
 367 tivity k , the bedrock drainable porosity ϕ_d and the bedrock total porosity ϕ_{tot} . Accord-
 368 ing to the available data, our strategy is first to select the most promising flow models
 369 defined by their hydraulic conductivity k and their drainable porosity ϕ_d with the monthly
 370 discharge observations and second to determine the total aquifer porosity ϕ_{tot} with the
 371 sampled CFCs measurements.

372 Flows are modeled by the methods given in section 2.1 with a finely discretized mesh
 373 of 500 cells (4m wide). We calibrated the flow model parameters, i. e. the aquifer hy-
 374 draulic conductivity k and the drainable porosity ϕ_d , on the monthly discharge time se-
 375 ries. We ran 25,000 simulations with a Monte Carlo sampling method, k in the range
 376 $[10^{-8}, 10^{-4}]$ m/s and ϕ_d in the range $[0.005, 0.5]$. The goodness of fit is assessed by the
 377 Kling-Gupta efficiency (KGE) (Gupta et al., 2009). KGE increases for better models and
 378 reaches 1 for a perfect fit.

379 We selected the 39 best simulations giving the highest KGE values and ran the advective-
 380 dispersive transport on them for different ϕ_{tot} values in $[\phi_d, 0.5]$. This resulted in 147
 381 transport simulations on which CFC-11 and CFC-113 concentrations ($C_{CFC}^{mod}(t_s, x)$ ex-
 382 pressed in equivalent to atmospheric concentration, where t_s is the sampling time and
 383 x is the distance of the sampling location to the river) are estimated in the river and in
 384 the wells, under the assumption of no atmospheric contamination, with the classic con-
 385 volution equation (Marçais et al., 2015). We then calibrated the hillslope transport model
 386 on the CFC concentrations, by minimizing the following objective function:

$$\begin{aligned}
 \Theta(k, \phi_d, \phi_{tot}) = & 1 - \frac{1}{6} \left(\frac{\sum_{i=1}^3 |C_{CFC11_i}^{mod}(x_i) - C_{CFC11_i}^{obs}(x_i)|}{\sum_{i=1}^3 |C_{CFC11_i}^{obs}(x_i)|} \right. \\
 & \left. + \frac{\sum_{i=1}^3 |C_{CFC113_i}^{mod}(x_i) - C_{CFC113_i}^{obs}(x_i)|}{\sum_{i=1}^3 |C_{CFC113_i}^{obs}(x_i)|} \right) \quad (6)
 \end{aligned}$$

388 Like the *KGE* indicator, Θ gets close to 1 for perfect fits.

389 3 Results

390 We present the calibration results and the capacity of the discharge and CFCs data
 391 to constrain the aquifer hydraulic parameters (section 3.1). The calibrated models show
 392 the control of the streamwater mean transit times by the groundwater contribution (sec-
 393 tion 3.2) and illustrate its seasonal evolution with the dynamic partitioning of the over-
 394 land flow, seepage flow and baseflow (section 3.3).

395 3.1 Calibration of the aquifer hydraulic parameters

396 3.1.1 Calibration of the hydraulic conductivity by discharge data

397 Hydraulic conductivity k and drainable porosity ϕ_d are evaluated by comparing the
 398 observed and simulated discharge data at the Guillec catchment outlet according to the
 399 *KGE* metric (Figures 4.A, S4, S5 and Text S4). The modeled discharge is highly sen-
 400 sitive to both k and ϕ_d with *KGE* values broadly ranging between 0.040 and 0.82. The
 401 local maximum values of *KGE* are underlined by the dashed red curve and cover a wide

402 range of k and ϕ_d values organized around two trends respectively horizontal and ver-
 403 tical (Figures 4.A, S6 and Text S5). For the vertical trend, k is well constrained around
 404 10^{-5} m/s while ϕ_d broadly varies between 2 % and 47 %. For the horizontal trend, ϕ_d
 405 is smaller around 1 % while k varies between 10^{-5} m/s and 9×10^{-5} m/s. It shows that
 406 discharge data constrain the hydraulic conductivities within less than one order of mag-
 407 nitude (from 1×10^{-5} m/s to 9×10^{-5} m/s) out of the 4 orders of magnitude explored
 408 (from 10^{-8} m/s to 10^{-4} m/s). The drainable porosity ϕ_d is however poorly constrained
 409 with the best simulations spanning almost all the explored range, from 0.6 % to 47 %.
 410 At this stage, We select the 39 sets of parameters (k, ϕ_d) giving the best KGE values (i.e.
 411 $KGE \geq 0.8$, indicated by dots on Figure 4.A).

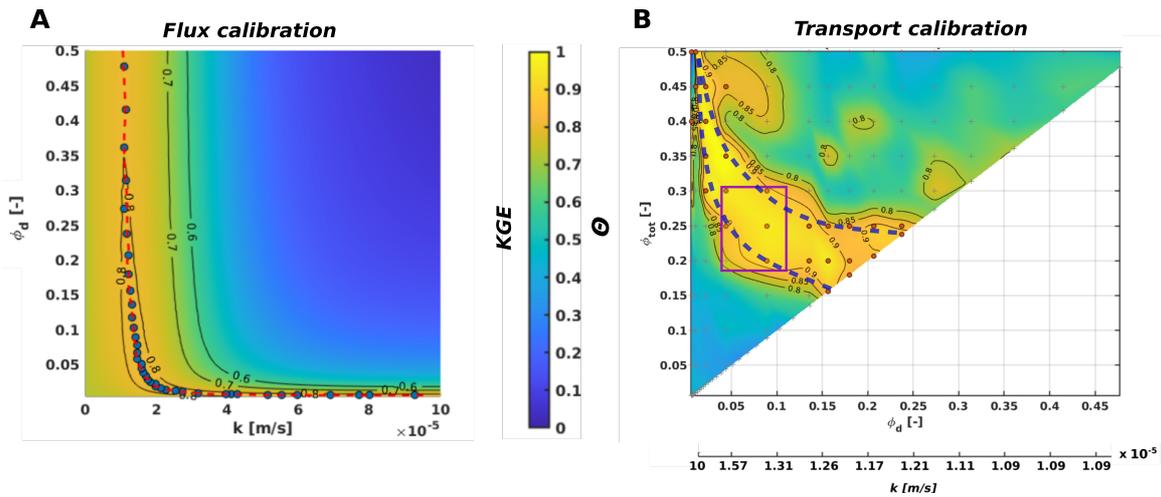


Figure 4. A. KGE calibration performance on discharge data displayed for the 25 000 simulations sampling the (k, ϕ_d) parameter space. The dotted red curve indicates the location of the local maxima joining the 39 best models retained for the analysis (blue dots). B. Θ calibration performance on CFCs data (6) in the (ϕ_d, ϕ_{tot}) space. Orange circles represent the 28 best transport models. Dashed blue lines materialize the constrained zone in the (ϕ_d, ϕ_{tot}) map. The purple rectangle further restrains the acceptable values within the range of values previously reported in the literature for the same type of lithologies (section 4.1). In the right figure, values of the hydraulic conductivity k corresponding to the drainable porosity ϕ_d (obtained from the calibration of the discharge data, section 3.1) are represented on the lower scale (scale is not linear).

412 3.1.2 Calibration of drainable and total porosities by CFCs data

413 Starting from the 39 best flow models obtained with the previously identified $(k,$
 414 $\phi_d)$ sets, we defined 147 transport models with different values of ϕ_{tot} to explore the pa-
 415 rameter space defined by $\phi_{tot} \geq \phi_d$. We assessed the agreement between modeled and
 416 sampled CFC concentrations with the function Θ (equation 6 and Figure S7). Results
 417 show that a close match is obtained for 28 of the 147 models with values of Θ greater
 418 than 0.8 defining a characteristic zone of acceptance in the (ϕ_d, ϕ_{tot}) space (Figure 4.B).
 419 By construction, the 28 models match both the sampled CFC concentrations and the
 420 seasonal discharge variations.

421 ϕ_d and ϕ_{tot} are partly constrained by CFC data (dashed blue lines on Figure 4.B),
 422 while they were not by the discharge data. Admissible ϕ_d and ϕ_{tot} lead to streamwater
 423 mean transit times equal to 13 ± 2 years (Figures 5.B, S8 and S9). Any decrease of the
 424 drainable porosity ϕ_d is balanced by some increase of the total porosity ϕ_{tot} to compen-
 425 sate the speed up of the drainable flowpaths by some additional delay in the immobile
 426 porosity ($\phi_{tot} - \phi_d$). ϕ_d and ϕ_{tot} do not however have exactly the same effect. ϕ_d mod-
 427 ifies the speed of the drainable flowpaths, the dynamical storage of the aquifer and its
 428 seasonal amplitude as well as the water table intersection with the topography and its
 429 dynamic. In contrast, ϕ_{tot} more directly scales the transit times (equation 3) through
 430 a simple linear dependency, which clearly appears at steady state (Text S6 and Figure
 431 S8).

432 3.2 Controls of streamwater mean transit times by groundwater con- 433 tributions

434 The contribution of groundwater to the streams is characterized by the Ground-
 435 water Index. The Groundwater Index, abbreviated GWI, is defined as the proportion
 436 of baseflow and seepage flow to the streamflow. It is on average equal to 75 % with a lim-
 437 ited variability among the 39 best models (± 14 %, Figure 5.A). When it reaches the stream,
 438 groundwater is on average 17 ± 3.5 years old (Figure 5.B). As a result, streamwaters are
 439 relatively old with a mean transit time of 13 years ($\approx 17 \times 0.75$) assuming that over-
 440 land flows are less than 1 year old and intervene mostly through a dilution effect. Even
 441 in the wet season when its contribution is reduced, groundwater still amounts to 40 %
 442 of the stream discharge, explaining why streamwater transit times remain "old" (≈ 6.6
 443 years old).

444 Resulting from the Boussinesq response and from the seasonal water table inter-
 445 section with the topography, GWI, the groundwater contributions to the stream varies
 446 significantly throughout the year. GWI decreases from 0.98 at early low flows (August),
 447 when streams are almost exclusively fed by aquifers, to 0.39 at early high flows (Decem-
 448 ber), when precipitation and saturated areas are maximal (Figure 5.A). The streamwa-
 449 ter mean transit time (mTT) displays similar variations with a time lag of 1-2 months,
 450 from 20 years at late low flows (October) to 6.5 years at the late high flow season (Febru-
 451 ary) (Figure 5.B). The mean transit time of the groundwater contribution to the stream
 452 (mTT_{gw}) is further delayed by one month ($mTT_{gw} \in [14.5, 33]$ years). This is because
 453 deeper compartments of the catchment have a delayed response compared to the shal-
 454 lower compartments. mTT_{gw} also displays an increased variability between the calibrated
 455 models (± 3.5 years for mTT_{gw} to compare with ± 1.7 years for the mTT).

456 This shows that the groundwater contribution characterized by GWI, which is con-
 457 trolled by the aquifer hydraulic conductivity k (section 3.1.1), itself controls the seasonal
 458 dynamics of streamwater mean transit times mTT . Below, we further investigate the ef-
 459 fect of flowpath decomposition on the full transit time distribution (TTD) and its sea-
 460 sonal variations.

461 3.3 Structure of the Transit Time Distributions as revealed by the flow- 462 path organization

463 The decomposition of the transit time distribution in these three flowpath compo-
 464 nents (overland flow, seepage flow and baseflow) shows how the spatial stratification of

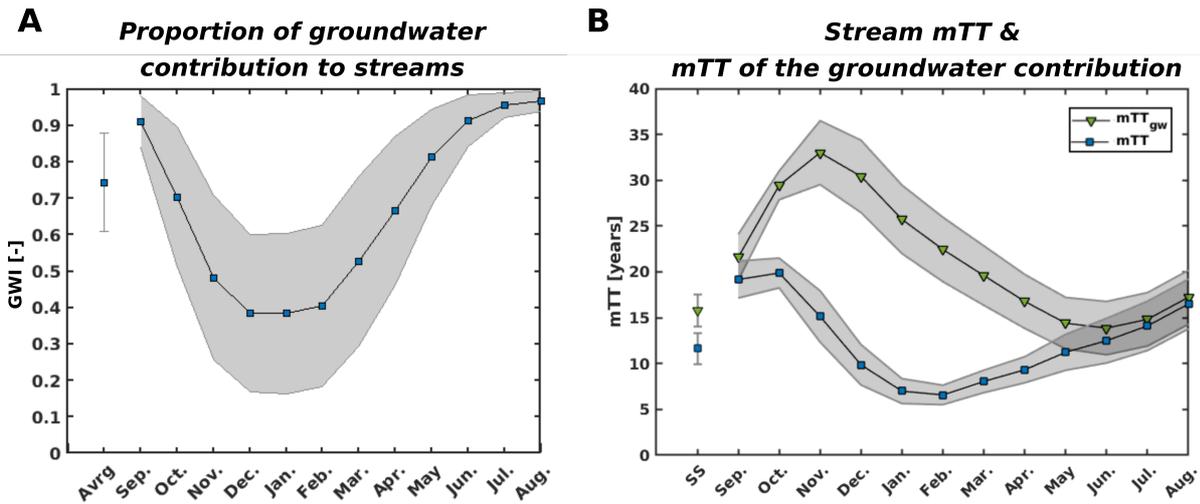


Figure 5. Modeled interannual monthly variations of A. the proportion of groundwater contributions to stream characterized by the Groundwater Index GWI and B. the streamwater mean transit times (mTT , squares) and mean transit times of the groundwater contribution to stream (mTT_{gw} , triangles). Note how the contribution of the aquifer is delayed compared to the river signal. The maximum mTT_{gw} is reached in November, one month later than mTT (maximum attained in October). The shaded area marks the standard deviation obtained with the 28 best transport models (out of 147 transport models). First points and error bars on the left labeled "Avg" and "SS" stand respectively for the annual average and for the steady state value.

465 the flowpaths is translated in the time domain both in low flow and high flow periods
 466 (Figure 6.A). Overland flows have the fastest times centered on one month (0.1 year) (red
 467 curves). Seepage flows have an intermediate timescale centered on one year (green curves).
 468 This timescale of one year reflects the seasonal seepage front development at the top of
 469 the aquifer composed of groundwater with a residence time less than one year since it
 470 is flushed every year. Third, the baseflow contribution has the oldest transit times with
 471 an average timescale of 10 to 30 years depending on the season (blue curves).

472 Differences in the TTD decomposition between seasons (at high flow vs low flow)
 473 lead to some counterintuitive results. At high flows (curves indicated by diamonds), the
 474 transit times of the overland and seepage flows are surprisingly slightly longer than at
 475 low flows. It results from the development of the seepage zone and reflects the fact that
 476 runoff occurs from farther away from the river on the hillslope increasing the mean transit
 477 time of overland flowpaths to the river. With receding seepage (crosses), flowpaths
 478 emerging in the seepage zone become shorter by occurring closer to the stream. Similar
 479 conclusions have been drawn by Wilusz et al. (2020). More intuitively, baseflow con-
 480 tribution is older at low flows as it comes from deeper older compartments due to the
 481 groundwater age stratification. Interestingly, the transit time distributions of the base-
 482 flow contribution is the most sensitive to seasonal variations with transit timescales sig-
 483 nificantly shifted from high flow (10 years) to low flow (30 years), in line with the mTT_{gw}
 484 seasonal variations (triangles in Figure 5.B).

485 However, we observe that the large seasonal mean transit time variations (squares
 486 in Figure 5.B) do not result from the intrinsic mTT variability of the different flowpath

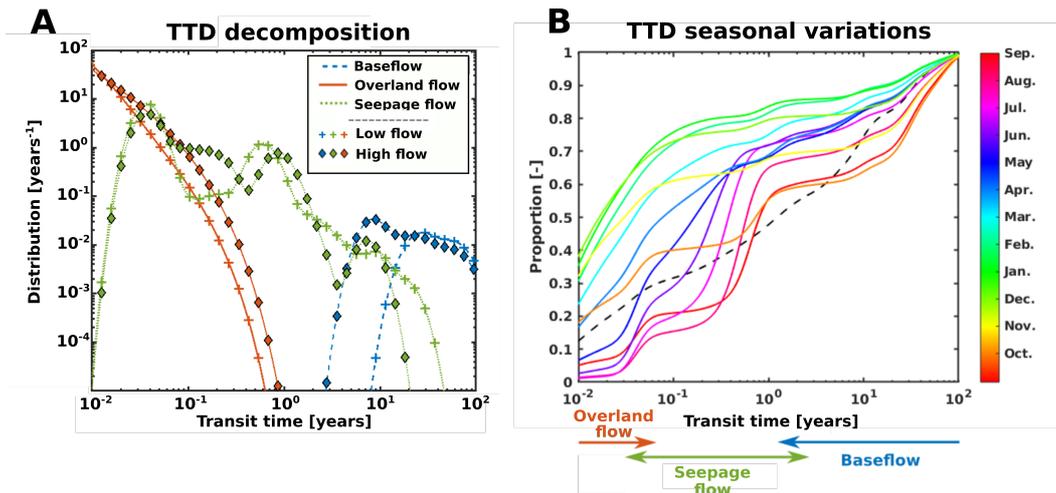


Figure 6. A. TTD decomposition in the streamflow components illustrated by Figure 1: overland flow (red solid lines), seepage flow (green dotted lines) and baseflow (blue dashed lines). TTDs are given at low flows in September 2016 (crosses) and at high flows in February 2017 (diamonds). B. Interannual averaged seasonal cumulative TTDs. During the wet season, the proportion of young water in the river increases significantly. The black dotted line represents the steady state transit time distribution. Transport simulation were obtained for $k = 1.6 \times 10^{-5}$ m/s, $\phi_d = 4\%$ and $\phi_{tot} = 25\%$. Note the well-separated characteristics transit times of the three flow components on both figures.

487 components (overland flow, seepage flow or deep flow), but rather from the time-varying
 488 respective contribution of these fluxes to the stream discharge, in accordance with pre-
 489 vious studies carried with the SAS framework (Rodriguez & Klaus, 2019; Wilusz et al.,
 490 2020). The seasonal variations of the cumulative transit time distributions shown by Fig-
 491 ure 6.B confirm this. On this plot, the lines of the cumulative TTDs indeed display sharp
 492 increases at specific transit timescales illustrating shifts in the relative proportion of these
 493 flowpaths.

494 4 Discussion

495 4.1 Complementarity of discharge and CFCs data informs catchment 496 scale hydrological properties and processes

497 Discharge data turn out to illuminate hydraulic conductivity at the catchment scale
 498 by constraining its range to less than one order of magnitude between 10^{-5} m/s and $9 \times$
 499 10^{-5} m/s. Such values of hydraulic conductivities are consistent with previous ranges
 500 of hydraulic conductivities and transmissivities obtained at comparable scales for the crys-
 501 talline basements of Brittany as shown by Table 1 (Conan et al., 2003; Grimaldi et al.,
 502 2009; Molenat, 1999; Clément et al., 2003; Legchenko et al., 2004; Martin et al., 2006;
 503 Kolbe et al., 2016; Leray et al., 2012; Le Borgne et al., 2006; Roques et al., 2014). The
 504 limited range of estimated hydraulic conductivity and the agreement with independently
 505 derived values show that this method adequately constrains hydraulic conductivities at
 506 the catchment scale. It offers a complementary approach to the classic analysis of reces-
 507 sion first introduced by Brutsaert and Nieber (1977), which requires essential but difficult-

508 to-assess assumptions for extracting the baseflow contribution (Wittenberg & Sivapalan,
509 1999; Troch et al., 2013; Vannier et al., 2014).

510 Fundamentally, the limited range inferred for the hydraulic conductivity comes from
511 the sensitivity of the partitioning process between overland flows and groundwater flows
512 in the discharge data. The contribution of baseflow and seepage flows sustaining the river
513 discharge, estimated on average at 75 %, is indeed consistent with previous first-order
514 estimates of around 60 % (Mougin et al., 2008; Aquilina et al., 2012; Clément et al., 2003)
515 as well as with statistical filter estimates of around 79 % (Gustard et al., 1992; Lyne &
516 Hollick, 1979). The contribution of overland flows is directly controlled by the distribu-
517 tion of saturation along the hillslope (Kosugi et al., 2008; Montgomery et al., 1997; Wil-
518 son & Dietrich, 1987; S. P. Anderson et al., 1997), which is set by the water table po-
519 sition relative to the surface. At first order, this position of the water table is controlled
520 by the hydraulic conductivity and not by the drainable porosity. This is intuitively il-
521 lustrated by the absence of dependence of the steady-state groundwater flow equation
522 on the drainable porosity. The drainable porosity only intervenes as a second-order con-
523 trol for the typical seasonal dynamics of the water table and the subsequent seasonal base-
524 flow contributions to the streamflow (Figure S8). This effectively explains why the drain-
525 able porosity range cannot be constrained by the discharge datasets alone (Guérin et al.,
526 2019). It also explains why the overall model uncertainty is not strongly impacted by
527 the uncertainty of the drainable porosity.

k [m/s]	T [m ² /s]	ϕ_d [-]	ϕ_{tot} [-]	Depth [m]	N	Settings	Reference
$1.8 \cdot 10^{-5}$	$3.1 \cdot 10^{-4}$	0.080	0.29	17	28	Guillec: shallow granites	This study
$8.0 \cdot 10^{-6}$		0.080	0.60	20	12	Kerrien, Kerbernez: shallow granites	Martin et al. (2006)
$3.6 \cdot 10^{-5}$		0.050		30	10	Kerrien, Kerbernez: shallow granites	Legchenko et al. (2004)
$1.0 \cdot 10^{-5}$				10	2	Pleine Fougères: weathered schists	Clément et al. (2003)
$6.3 \cdot 10^{-6}$			0.45	20	10	Pleine Fougères: weathered schists and granites	Kolbe et al. (2016)
$2.6 \cdot 10^{-6}$			0.21	50	10	Pleine Fougères: fractured schists and granites	Kolbe et al. (2016)
$5.1 \cdot 10^{-6}$		0.030		6	7	Kervidy: weathered schists	Molenat (1999)
$2.2 \cdot 10^{-6}$					4	Pont Lagot: weathered schists	Grimaldi et al. (2009)
$1.0 \cdot 10^{-5}$		0.013			2	Kervidy: weathered schists	Conan et al. (2003)
	$1.3 \cdot 10^{-5}$				2	Kervidy: fractured schists	Conan et al. (2003)
	$2.2 \cdot 10^{-3}$	0.037			11	Ploëmeur: transmissive zone, schists and granites	Leray et al. (2012)
	$6.0 \cdot 10^{-4}$			≤ 25	12	Ploëmeur: shallow schists	Le Borgne et al. (2006)
	$2.2 \cdot 10^{-4}$			≥ 25	22	Ploëmeur: deep granites	Le Borgne et al. (2006)
	$2.9 \cdot 10^{-3}$				23	Ploëmeur: schists and granites	Le Borgne et al. (2006)
	$7.3 \cdot 10^{-5}$	0.035		15	4	St Brice: weathered schists	Roques et al. (2014)
	$4.4 \cdot 10^{-4}$			50	7	St Brice: fractured schists	Roques et al. (2014)

Table 1. Logarithmic averaged measured or modeled hydraulic properties in Brittany aquifers. For each site, independent estimates are provided as well as typical depth of investigation when available. N is the number of independent values leading to these estimations. For more details, see Text S7.

528 Drainable and total porosities can however be at least partially constrained by CFC
529 data. The drainable porosity ϕ_d must be between 0.6 % and 25 %, and the total poros-
530 ity ϕ_{tot} between 15 % and 47 % (Figure 4.B). Such ranges are consistent with previous
531 studies from which the drainable porosity inferred from well pumping tests or geochem-
532 ical characteristics has been found between 2 % and 10 % (Legchenko et al., 2004; Mar-
533 tin et al., 2006; Molenat, 1999; Leray et al., 2012; Roques et al., 2014) and the total poros-
534 ity inferred with CFC data between 25 % and 45 % (Kolbe et al., 2016) (see Table 1).
535 With these values, the porosities of the Guillec catchment would be further constrained

536 between 3 % and 12 % for ϕ_d and between 20 and 30 % for ϕ_{tot} (see purple rectangle
537 on Figure 4.B). The range of ϕ_{tot} might be quite high if not considering that the total
538 porosity is a lumped way to represent the different processes and structures that delay
539 the transport of solutes (Haggerty & Gorelick, 1995; Carrera et al., 1998).

540 4.2 The seasonal water table intersection with the land surface drives 541 the dynamics of the mean transit time mTT

542 The streamwater mean transit time mTT is equal to 13 years, almost twice as large
543 as the 7 years estimated previously on similar sites based on renewal rates (Mougin et
544 al., 2015). Even though it is controlled by the averaged size of the groundwater storage
545 like in other similar sites (Kolbe et al., 2016; Haitjema, 1995; Chesnaux et al., 2005), the
546 mean transit time differs from the renewal rate defined as the groundwater storage di-
547 vided by the recharge (Danckwerts, 1953). Two of the assumptions underlying the re-
548 newal rate approximation noted in Harman (2014) and Berghuijs and Kirchner (2017)
549 are not fulfilled. First, the modeled processes are inherently transient. Differences be-
550 tween the steady state (i.e. obtained when the state variable S does not vary) and the
551 stationary state (i.e. the average of the seasonal variations) mTT and mTT_{gw} are sig-
552 nificant (Table S2). This is also clearly illustrated on Figure 7, where the steady state
553 mean transit time of the groundwater contributions to the stream mTT_{gw} is significantly
554 shorter than the seasonal mTT_{gw} variations. This comes from the non linearity of the
555 model incorporating seepage areas that strongly depend on the precise water table lo-
556 cation. Detailed explanations are given in Text S8. Second, the reservoir is not perfectly
557 mixed. The transit time distribution is therefore rather power-law than exponential be-
558 cause the reservoir remains stratified with only partial mixing and is limited at the wa-
559 ter table by the seasonal intersection with the land surface.

560 These results confirm the measurements carried out regionally on several Brittany
561 watersheds and more globally in temperate catchments (J. Yang et al., 2018). From a
562 depth of a few meters below the water table, the residence time is greater than 10 years,
563 but conversely, the yearly-flushed, variably saturated zone shows residence times on the
564 order of one year (Ayraud et al., 2008; Legout et al., 2007; de Montety et al., 2018). The
565 results obtained here make it possible to remove a contradiction between the observa-
566 tions of residence times measured by tracers in groundwater which indicate long times
567 and the fast processes generally associated with streamwater discharge dynamics. Long
568 residence times in groundwater (through baseflow) together with a rapid component of
569 transfer into the rivers (through overland and seepage flow) are not incompatible, rather
570 they are linked through the seasonal water table fluctuations and groundwater strati-
571 fication.

572 We further explore how these modeled processes control the temporal dynamics of
573 the mean transit time mTT . To highlight their impact on the model results we perform
574 two complementary simulations where we alternatively remove water table intersection
575 and stratification from the reference model (squares in Figure 7). The reference model
576 refers to the model, which has been used in this study (described in section 2). First,
577 we removed the water table interactions with the land surface (stars in Figure 7) from
578 the reference model by not considering the upper land surface boundary on the aquifer
579 capacity (last equation of system of equations (1)). For this model, the seasonal dynam-
580 ics of mTT almost vanishes with variations strongly limited to 1.7 years (Figure 7.A).

581 Second, we removed the groundwater stratification (diamonds in Figure 7) from the ref-
 582 erence model. To do that, we relaxed the implicit condition stated in the second equa-
 583 tion of system (3) that the particles exiting the first the aquifer at the seepage face are
 584 the ones located at the top of the aquifer column. Instead, we developed an algorithm
 585 that randomly samples particles, from any vertical location of a given numeric cell en-
 586 counter seepage face. The right number of particles is retrieved to feed the seepage
 587 flux by ensuring flux conservation. For this latter model, the seasonal dynamics of the
 588 mean transit time is well preserved as well as its magnitude. The remaining differences
 589 mainly occur during the dry season (± 4 years at most during the September month) when
 590 transit times are dominantly controlled by the proportion of groundwater contributions
 591 to streamflow.

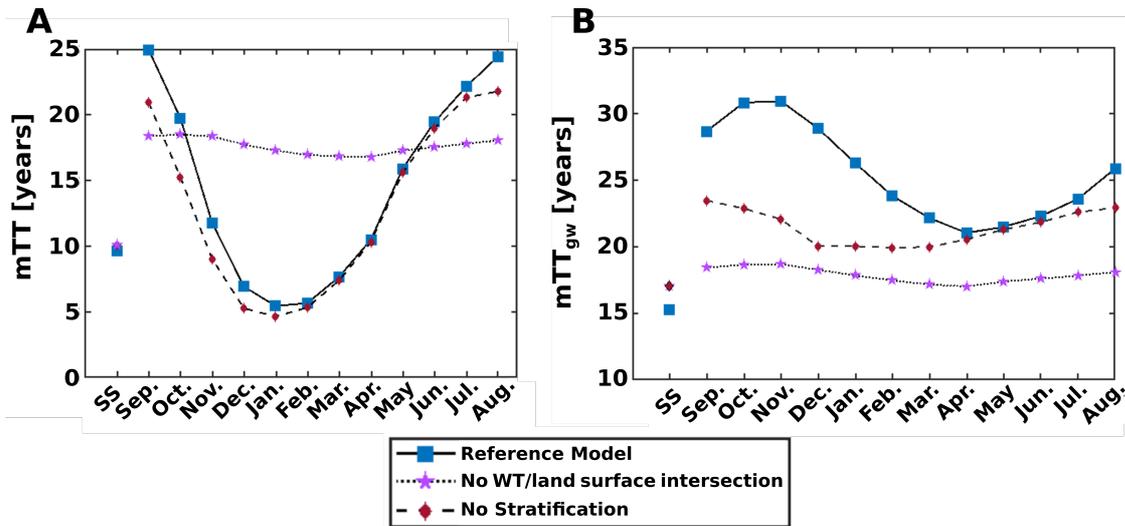


Figure 7. Investigation of the relative effects of the seasonal water table intersection with the land surface and of the groundwater flowpaths stratification for the steady state (SS) and interannual monthly averaged. Comparison of A. the mean transit time to the river mTT and B. the mean transit time of the groundwater contribution (baseflow and seepage flow) to the streamwater mTT_{gw} for the reference model (squares) and the models without intersection with the surface (stars) and without groundwater stratification (diamonds). The three models are parametrized with: $k = 1.17 \cdot 10^{-5}$ m/s and $\phi_d = \phi_{tot} = 24$ %. Note that the difference of vertical scales between the two figures.

592 Two compartments (here the surface and the aquifer), with different characteris-
 593 tic transit times, are therefore needed to obtain seasonally varying mTT . Their dynam-
 594 ical connection through water table intersection indeed generates temporary saturation
 595 excess overland flow and seepage flow leading to seasonally varying mTT . With only the
 596 groundwater compartment, the seasonally varying infiltration fluxes are dampened by
 597 the groundwater storage volume and mTT becomes nearly constant. Groundwater strat-
 598 ification alone has a notably smaller impact on the streamflow mTT since mTT vari-
 599 ations are primarily controlled by the relative contribution of groundwater flows to the
 600 stream and not by the internal organization of groundwater circulation inside the aquifer.

601 Here the short streamwater transit times come from the contribution of the sur-
602 face compartment to the stream. While the groundwater compartment is key to explain
603 the old transit times (≥ 1 year old), faster flowpaths can be delivered by the surface or
604 by the soil compartment. Deciphering the relative role of these shallower compartments
605 would require other tracers than CFCs and a model running with finer timesteps (e.g.
606 hourly to daily timesteps).

607 **4.3 Groundwater flowpath stratification is critical for the mean tran-** 608 **sit time of the groundwater contribution to the river mTT_{gw}**

609 While groundwater stratification is a second-order control of the mean transit time
610 dynamics of the entire streamflow, it becomes essential for the mean transit time dynam-
611 ics of the groundwater contribution to the river, mTT_{gw} (Figure 7.B). The simplified model
612 without stratification (diamonds) misses the seasonal dynamics of the reference hillslope
613 model (squares). Indeed, for the reference model, mTT_{gw} broadly varies by ± 4 years around
614 a mean of 27 years. On the contrary, the model without groundwater stratification only
615 varies by ± 1.5 years around a mean of 23 years. Randomly sampling the particles from
616 the water column indeed removes the preference for the aquifer to release young water.
617 On the contrary, the reference model, which respects the stratification of groundwater
618 residence times, mostly samples young groundwater during the wet season, when such
619 young particles are available at the top of the water column. During the dry season, only
620 old groundwater particles remain to be sampled, reinforcing the seasonal variability of
621 mTT_{gw} .

622 The quantity mTT_{gw} is likely to be crucial when a precise representation of deep
623 compartments is needed. This encompasses a broad range of topics, from studies assess-
624 ing the impact of global climate change on transit times and water quality to studies in-
625 terested in legacy effects of these groundwater storages, including geochemical processes
626 where the interaction of reaction kinetics and transit times can be important (White et
627 al., 1983). Both measured and model-derived water ages have been shown to correlate
628 with weathering-derived solute chemistry in streams (Benettin et al., 2015; Rademacher
629 et al., 2005). Geochemical tracer data such as Ge/Si ratios and stable isotopes of Si and
630 Ca indicate that not only the flux but also the sources of solutes vary with discharge in
631 studied streams (Kurtz et al., 2011; Cenko-Tok et al., 2009). The mechanism can include
632 interactions with different materials along different flowpaths, but also differences in the
633 time available for reaction progress as in kinetically limited systems (Fernandez et al.,
634 2021). Integrated hydro-biogeochemical models, like the approach outlined here, rely-
635 ing on partitioning flow and timing transport are therefore needed to address these hy-
636 potheses.

637 **4.4 Limitations and perspectives**

638 While this parsimonious strategy effectively constrains the transit times of the ground-
639 water circulation, some limitations arise from the conceptual representation of the sur-
640 face and unsaturated zone processes. First, surface runoff processes are represented through
641 a gravity-driven routing representation. This routing scheme neglects pre-event water
642 present in the soil matrix, which can contribute to the flow. The presence of pre-event
643 water could delay the estimated transit times of overland flowpaths by some months but
644 will not significantly change the estimate of the mean transit times and the old compo-

645 nent of the streamwater transit times distributions. However, this limitation prevents
646 us from applying the model in its current configuration at the daily to hourly timescale.
647 Indeed, we would strongly underestimate runoff routing transit times that have been shown
648 to be mainly made up of pre-event water (von Freyberg et al., 2018). Adding a more de-
649 tailed runoff component but still parsimonious to this equivalent hillslope representation
650 could be achieved through a total soil water storage description that could be partially
651 mixed with the runoff routing. This would also enable the model to distinguish between
652 the fast flowpaths arising at the surface or in the soil compartment (Heidbüchel et al.,
653 2020). Second, this routing scheme process neglects re-infiltration typically occurring be-
654 fore surface overland flow reaches the stream (Jackson et al., 2014; Klaus & Jackson, 2018;
655 Gabrielli & McDonnell, 2020). In our model, most of the saturation excess overland flow
656 is generated in the vicinity of the river. This proximity to the river might prevent re-
657 infiltration (Jackson et al., 2014). Nevertheless, if re-infiltration occurs here, increased
658 transit times would be less than one year as the groundwater transit times for water in-
659 filtrating in the vicinity of the river is less than one year (Figure S10).

660 Another limitation refers to the transfers in the unsaturated zone, which have been
661 neglected. Studies of transit times in the unsaturated zone in crystalline basement have
662 shown that they are short (ca. 2 years) (Legout et al., 2007; de Montety et al., 2018),
663 and so integration of an explicit representation of unsaturated zone transfer should have
664 marginal impact on the longer transit times associated with groundwater flow. Indeed,
665 in crystalline basements located in temperate climate, the unsaturated zone is quite shal-
666 low (typically less than 5 meters, (Le Borgne et al., 2006; Roques et al., 2014; Marçais
667 et al., 2018)). This could be different for other settings (arid climate, sedimentary set-
668 tings), where the representation of percolation in the unsaturated zone will be a signif-
669 icant part of the total transit times to the stream (Schwientek et al., 2009; Green et al.,
670 2018; Chen et al., 2019). Adding a flux and transport representation of the vadose zone
671 could be done with a 1D vertical model, which would preserve the parsimonious philos-
672 ophy of the approach presented here for the saturated compartment (Harman et al., 2011).

673 All these limitations would increase the young transit times, but would have only
674 small effects on the old transit times. Further, constraining these fast processes would
675 require the addition of specific tracers well-suited for short transit times. Indeed, CFCs
676 tracers cannot track transport in the unsaturated zone and are not adapted to the timescales
677 below one year, which require intensive monitoring of stable water isotopes (2H , $\delta^{18}O$)
678 or conservative solutes (chloride) (Sprenger et al., 2019; Harman, 2014; Benettin, Soulsby,
679 et al., 2017).

680 An additional limitation in this study is related to the number of CFCs samples
681 used to calibrate the model. We only relied on one sampling campaign on three differ-
682 ent sampling sites to date the groundwater and its contribution to the river. A solution
683 would be to sample more sites, e.g. by systematically sampling springs (Rademacher et
684 al., 2001, 2005) or to sample CFCs at different seasons (Guillaumot et al., 2021). More
685 intensive sampling indeed enables to link discharge (Morgenstern et al., 2010) or dissolved
686 solutes concentrations to groundwater ages (Marçais et al., 2018), which offers oppor-
687 tunities to strengthen the spatial and temporal resolution of transit times in the river
688 (Peters et al., 2014).

5 Conclusion

Using a parsimonious approach that emphasizes 1) variable contributions from shallow and deep flow paths and 2) stratification of groundwater residence times, we have demonstrated how the groundwater contribution to streams control streamwater mean transit times and its seasonal variations in a crystalline bedrock catchment under temperate climate. The 2D model requires the calibration of the hydraulic conductivity, drainable porosity and total porosity from discharge time series and CFC data. In the catchment studied, all the calibrated models display significant groundwater contribution to stream (ca. $75\pm 14\%$) with an average age of 17 years, leading to mean streamwater transit times on average equal to 13 years. All calibrated models display strong seasonal interactions between the water table and the land surface leading to a dynamic partitioning between groundwater flows (baseflow + seepage flow) and overland flows. This partitioning strongly conditions the seasonal dynamics of the streamwater mean transit times, from 6 years at high flow to 20 years at low flow.

The influence of the groundwater residence time stratification plays only a second-order role for the seasonal evolution of streamwater mean transit times, but is significant for the seasonal evolution of the mean transit times of the groundwater contribution to the stream. This last finding is likely to play a pivotal role for compounds which are predominantly found or generated in the aquifer. These may include solutes released from bedrock weathering or contaminants transported in groundwater.

The hillslope models show how (1) surface vs groundwater flow partitioning and (2) groundwater residence time stratification shape the transit time distributions. The understanding obtained through this study is particularly important for issues related to diffuse pollutant transfers and more generally to element transfers and biogeochemical fluxes within catchments. It is common for workers in the field to focus on in-stream processes and rapid dynamics that impact river flow, such as overland flow. Here we show that most of the streamwater has in fact resided underground for time scales of a decade or more and is delivered to the stream as baseflow and seepage flow. The decomposition of the flow sources also shows how the transfer of elements via groundwater is distributed between a fast component (one year for the seepage flow) and a much slower component (several tens of years for the baseflow), which indicates that changes in agricultural practices or land use will both have a rapid impact (ca. 1 year) and will also have impacts over the long term (tens of years). This complexity should be accounted for in watershed management.

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732 **Appendix A Separation of lateral and vertical velocity term**

733 Considering that the water is incompressible, we have the relation $\nabla \mathbf{v} = \frac{dv_x}{dx} +$
 734 $\frac{dv_y}{dy} + \frac{dv_z}{dz} = 0$ (Strack, 1984; Pollock, 1988; Harman, 2019). Deriving v_x with respect
 735 to x (cf. equation 2), we get:

$$736 \begin{aligned} \frac{dv_x}{dx} &= \frac{\phi_d}{\phi_{tot}} v_x \left(\frac{1}{Q} \frac{dQ}{dx} - \frac{1}{S} \frac{dS}{dx} \right) \\ &= \frac{\phi_d}{\phi_{tot}} v_x \left(\frac{1}{Q} \frac{dQ}{dx} - \frac{1}{h} \frac{dh}{dx} - \frac{1}{\omega} \frac{d\omega}{dx} \right) \end{aligned} \quad (7)$$

737 Applying symmetry principles (Curie, 1894), we can attribute the part of this quan-
 738 tity invariant to z to the quantity $\frac{dv_y}{dy}$:

$$739 \frac{dv_y}{dy} = \frac{\phi_d}{\phi_{tot}} \frac{1}{\omega} \frac{d\omega}{dx} v_x \quad (8)$$

740 This enables derivation of v_z :

$$741 \begin{aligned} \frac{dv_z}{dz} &= -\frac{dv_x}{dx} - \frac{dv_y}{dy} \\ &= -\frac{\phi_d}{\phi_{tot}} \left(\frac{d}{dx} \left(\frac{Q}{S} \right) + \frac{1}{\omega} \frac{d\omega}{dx} \frac{Q}{S} \right) \end{aligned} \quad (9)$$

742 Hence, as $\frac{dv_z}{dz}$ does not depend on z :

$$743 v_z = z \frac{dv_z}{dz} = -z \frac{\phi_d}{\phi_{tot}} \left(\frac{d}{dx} \left(\frac{Q}{S} \right) + \frac{1}{\omega} \frac{d\omega}{dx} \frac{Q}{S} \right) \quad (10)$$

744 2D flow lines $\mathbf{x}(t) = (x(t), z(t))$ derive from the Lagrangian definition of velocity where
 745 the temporal derivative of the trajectory $\mathbf{x}(t)$ is linked to the velocity field: $\frac{d\mathbf{x}}{dt} = \mathbf{v}(\mathbf{x}(t)) =$
 746 $(v_x(x(t), z(t), t), v_z(x(t), z(t), t))$.

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