



Environmental and hydrologic controls on sediment and organic carbon export from a subalpine catchment: insights from a time-series

Melissa S. Schwab^{1,2}, Hannah Gies¹, Chantal V. Freymond^{1,3}, Maarten Lupker^{1,4}, Negar Haghipour^{1,5}, Timothy I. Eglinton¹

¹Department of Earth Sciences, ETH Zurich, Zurich, 8092, Switzerland
 ²Now at Jet Propulsion Laboratory, California Institute of Technology, Pasadena, 91109, USA
 ³Now at Gruner, Basel, 4020, Switzerland
 ⁴Now independent, Bern, 3014, Switzerland
 ⁵Laboratory of Ion Beam Physics, ETH Zurich, Zurich, 8093, Switzerland

10 Correspondence to: Melissa S. Schwab (melissa.s.schwab@jpl.nasa.gov)

Abstract. Studies engaging in tracking headwater carbon signatures downstream remain sparse, despite their importance for constraining transfer and transformation pathways of organic carbon (OC) and developing regional-scale perspectives on mechanisms influencing the balance between remineralization and carbon export. Based on a 40-month time series, we investigate the dependence of hydrology and seasonality on the discharge of sediment and OC in a small Swiss subalpine

- 15 watershed (Sihl River basin). We analyze concentrations and isotopic compositions (δ^{13} C, F¹⁴C) of particulate OC and use dual-isotope mixing and machine learning frameworks to characterize and estimate source contributions, transport pathways, and export fluxes. The majority of transferred OC is sourced from plant biomass and soil material. Relative proportions of soil-derived particulate OC peak during the summer months, coinciding with maximum soil erosion rates. Bedrock-derived (petrogenic) OC abundant in headwater streams progressively decreases downstream in response to a lack of source material
- 20 and efficient overprinting with biospheric organic matter, illustrating rapid OC transformation over short distances. Large variations in isotopic compositions observed during baseflow conditions converge and form a homogenous mixture enriched in OC and characterized by higher POC- F¹⁴C values following precipitation-driven events. We propose that storms facilitate surface runoff and shallow landsliding, resulting in the entrainment of fresh litter and surficial soil layers. Model results further indicate diverging mobilization pathways. Discharge and water stage describe the export of suspended sediment,
- 25 while the prediction of POC fluxes is mostly supported by water stage and 1-day antecedent precipitation. Although particle transport in the Sihl River basin is mainly driven by hydrology, subtle changes in bedrock erosivity, slope angle, and floodplain extent likely have profound effects on the POC composition, age, and export yields.

1 Introduction

River networks serve as a continuum ultimately connecting the terrestrial with the marine biosphere (Aufdenkampe et al., 2011). Disproportional to their spatial extent, water bodies are active sites for transport, transformation, and storage of





significant portions of organic carbon (OC) mobilized from the terrestrial environment (Battin et al., 2009). Annually, between 1.90 and 2.95 PgC yr⁻¹ are entrained into inland waters (Cole et al., 2007; IPCC, 2021; Tranvik et al., 2009; Regnier et al., 2022). The majority of this carbon is lost during transfer due either to remineralization and outgassing or to burial in lakes and floodplains. Ultimately, only 0.80-0.95 PgC yr⁻¹ reach marine coastal regions (Tranvik et al., 2009; Battin et al.,

- 35 2009; Raymond et al., 2013; Lauerwald et al., 2015; IPCC, 2021; Regnier et al., 2022). However, anthropogenic and climate-driven changes markedly influence erosional processes and thus may perturb the translocation and sequestration of OC in freshwater systems. The human-induced lateral transfer of carbon adds ~0.60 PgC yr⁻¹ to inland waters which are largely respired and buried during fluvial transit (Regnier et al., 2022; Lauerwald et al., 2020; Li et al., 2019). The resulting OC flux to marginal shelves deviates by only 0.15 PgC yr⁻¹ from pre-industrial values (Regnier et al., 2022).
- 40 It is well established that small mountainous rivers deliver substantial quantities of sediment and particulate organic carbon (POC) to the oceans (Milliman and Syvitski, 1992; Lyons et al., 2002; Leithold et al., 2006; Hilton et al., 2012; Goñi et al., 2013). These systems are often characterized by a steep basin morphology, with little to no developed floodplains. The resulting basin storage capacity is insufficient to retain the large amounts of eroded sediments and soils exported from mountainous catchments (Wheatcroft et al., 2010; Milliman and Farnsworth, 2013). During storm-driven events,
- 45 mountainous rivers are strongly coupled to hillslope processes and often provide the main source of sediment to downstream channels (Milliman and Syvitski, 1992; Hilton et al., 2012).

A decrease in the OC content of the suspended load is commonly observed with increasing discharge in high sediment-yield small mountainous rivers (Masiello and Druffel, 2001; Coynel et al., 2005; Leithold et al., 2016). This decline corresponds to the dilution with carbon-poor bedrock material. Less well constrained is the character of the exported POC. The proportion of

- 50 fossil OC concomitantly increases with rising sediment yields in small river systems as they share a common source and transport pathways (Blair et al., 2003; Komada et al., 2004; Leithold et al., 2006; Hilton et al., 2011a). While deep-seated landslides and gully erosion mobilize predominantly bedrock-sourced (petrogenic) OC, surface runoff and shallow landslides preferentially remove fresh litter and organic-rich surface soils (Hovius et al., 2000; Hilton et al., 2008b; Hatten et al., 2012; Goñi et al., 2013).
- 55 In headwaters, significant portions of biospheric organic matter are exported in the form of coarse POC (>1 mm) encompassing leaves, needles, and wood fragments (Turowski et al., 2016; Rowland et al., 2017). In contrast to fine-grained OC, which can remain in suspension for prolonged periods, coarse particles are often deposited in headwater valley segments due to gravitational settling or retention in log jams (Wohl et al., 2012; Jochner et al., 2015). Interaction of woody debris and the gravel bedload might lead to grinding and size reduction, ultimately adding to the fine POC pool (Turowski et al., 2016).
- 60 Despite these mechanisms, restricting the transport of coarse POC, studies showed that event-driven floods can effectively recruit and transfer vascular plant debris as driftwood (West et al., 2011; Wohl and Ogden, 2013; Wohl, 2017; Ruiz-Villanueva et al., 2019) or as a component of the suspended load (Schwab et al., 2022) to continental margins.





Despite the growing recognition of the substantial and rapid downstream transfer of terrestrial OC from headwater streams (Leithold et al., 2016; Wheatcroft et al., 2010; Goñi et al., 2013) and of the processes controlling organic matter during its transfer through lowlands and floodplains (e.g., Bouchez et al., 2010; Hemingway et al., 2017; Repasch et al., 2021), river

- 65 transfer through lowlands and floodplains (e.g., Bouchez et al., 2010; Hemingway et al., 2017; Repasch et al., 2021), river segments connecting small, mountainous streams with lowland system remain poorly explored. A few studies address temporal dynamics of POC export in moderately steep river basins spanning timescales of individual storm events to intra- and inter- annual variability (Smith et al., 2013; Hatten et al., 2012). Even fewer studies examine the downstream evolution, composition, and molecular signature of OC in dynamic mountainous river systems (Goñi et al., 2014).
- 70 The focus of this study is to investigate in detail the response of sediment and bulk OC to variability in seasonality and discharge behavior in a moderately steep river basin bridging the gap between headwater streams and lowland rivers. The sub-alpine Sihl River links small mountainous headwater streams with the higher-order river Limmat, providing a crucial window on downstream transport and the evolution of OC along the riverine continuum. We obtained a high-resolution time series over 40 months focusing on the content, composition (δ¹³C, F¹⁴C), and flux of sediment and POC. Export fluxes are
- 75 modeled using traditional and machine learning approaches, while a dual isotope model framework allows the estimation of potential organic matter source contributions. We discuss control mechanisms regulating organic matter mobilization and transport and examine our results in the context of previously published data on Sihl River headwater catchments in order to derive comparisons regarding the nature of exported POC (Smith et al., 2013; Turowski et al., 2016; Gies et al., 2022).

2 Methods

80 2.1 Characteristics of the Sihl River watershed

The Sihl River basin located in the Swiss Prealps is part of the Rhine River headwater system (4th order tributary; **Fig. 1**). Its watershed covers an area of 346.0 km² ranging from an elevation of 1872 m (Druesberg) at its headwaters to 402 m at the catchment mouth, with an average slope of 19.5° (**Fig. 1a-b, Table 1**). The Sihl River basin experiences a humid continental climate with a mean annual air temperature of ~9.5°C and annual precipitation varying from 1450 to 1830 mm, with snowfall

- 85 generally occurring between November and April (MeteoSwiss, https://gate.meteoswiss.ch). The sub-alpine catchment is divided into an upper (145.7 km²) and lower basin (190.1 km²) near Einsiedeln by the reservoir Lake Sihl (10.2 km²) constructed in 1937 (Addor et al., 2011). The damming of Lake Sihl results in the abrupt fragmentation of the flow path and the effective capture of sediment entrained from the upper watershed. Further regularization structures such as weirs and the partly channelization with concrete embankments strongly diminish fluvial connectivity supporting the retention and
- 90 trapping of 62-67% of sediment in the lower Sihl River basin (Grill et al., 2019). The Alp and Biber River tributaries are free-flowing rivers and major sources of water and sediment to the Sihl River. Similar to other small mountainous river systems (Wheatcroft et al., 2010), the steep morphology of the watershed and the absence of extensive floodplains limits the water storage capacity of the Sihl River. In response to severe storms, discharges can rise from an average of 6.8 m³ s⁻¹ to



95



over 200 m³ s⁻¹ and result in devastating flash floods (e.g., August 2005: 280 m³ s⁻¹; Bezzola and Hegg, 2007; Jaun and Ahrens, 2009).

The land cover of the Sihl River basin in 2017 consisted of 38.9% meadows and pastures, 9.0% urban settlements, 6.1% unproductive areas, 1.3% water bodies, and 1.1% cropland (**Fig. 1c**; Federal Statistical Office, https://www.bfs.admin.ch), with the majority of the watershed covered by forests (43.4%; Waser et al., 2017). The main tree species in the upper basin are spruce (*Picea abies*) and fir (*Abies alba*) which are gradually replaced by deciduous trees such as beech (*Fagus*)

sylvatica), maple (Acer spp.), ash (Fraxinus spp.), and oak (Quercus petraea) in the lower basin (Schleppi et al., 2006).
 Forest soil carbon projections estimate stocks between 5 to 20 kgC m⁻² (Nussbaum et al., 2014; van der Voort et al., 2019).

The lithology of the lower Sihl River basin is composed of weakly consolidated, clastic sediments of the Molasse basin (**Fig. 1d**; Swisstopo; https://www.swisstopo.admin.ch). Limestones and other biogenic sedimentary rocks are common in the allochthon nappes, Klippen, and Säntis zones. Campanian-Maastrichtian and Eocene flysch (Schlieren and Wägital flysch),

105 largely composed of mudstone and calcareous sandstone, predominate in the Alp and upper Sihl River basins (Winkler et al., 1985). Dolomitic rocks of the northern limestone Alps and metamorphic rocks of the Arosa zone outcrop in the southern region of the upper Sihl River basin. Stream valleys and lowlands are filled with unconsolidated rock material such as alluvions, moraines, and gravel deposits that constitute important groundwater aquifers (Doppler et al., 2007).

The Erlenbach, Lümpenbach, and Vogelbach (Fig. 1a) are monitored as experimental catchments by the Swiss Federal

- 110 Institute for Forest, Snow, and Landscape (WSL), and are well studied in terms of terrestrial OC sources, mobilization, and export of fine and coarse material (Schleppi et al., 1998; Hagedorn et al., 2001; Turowski et al., 2011, 2016; Rickenmann et al., 2012; Smith et al., 2013; Gies et al., 2022; Hilton et al., 2021). These streams are second-order tributaries to the Sihl River, with catchment sizes ranging from 0.7 to 1.6 km² and average discharges of 38 to 77 L s⁻¹ (Smith et al., 2013; Gies et al., 2022). The Erlenbach basin is developed on an extensive bedrock landslide consisting primarily of Eocene Wägital
- 115 flysch (Winkler et al., 1985; Schuerch et al., 2006; Golly et al., 2017), while bedrocks in the Lümpenbach and Vogelbach are largely composed of calcareous sandstones (Milzow et al., 2006). The land cover of the drainage basins consists of alpine meadows, forests, and wetlands (Turowski et al., 2009; Gies et al., 2022).

2.2 Sample collection

From May 2014 to February 2015, surface water was sampled and processed by C. Freymond and H. Gies. Sample

120 collection from August 2016 to March 2019 was designed to capture variations in OC export in response to both seasonal changes and shorter-term variations in discharge behavior. We collected surface water samples from the Sihl River (Allmend Park, 47.35° N, 8.52° E; Fig. 1a) in a biweekly rhythm using a river-rinsed bucket. In addition, river water was collected during 17 storm events, emphasizing discharges >20 m³ s⁻¹. Although the water level can rise ~1.7 m during exceptional flood events, the Sihl River is generally characterized by water depths <1 m suggesting little vertical variations in suspended</p>





125 sediment and POC concentrations in the water column. Surface waters were retrieved seasonally from Lake Sihl (Sihl River inflow and two locations in the center of the lake) and the Alp River (Fig. 1a).

Known volumes of surface water (0.95 to 64.61 L) were filtered through pre-weighed and combusted (450°C, 6 h) 90 or 142 mm glass microfiber filters (GF/F, Whatman) with a nominal pore size of 0.7 µm using a steel filtration unit. The filtration occurred either in the field or immediately upon returning to the laboratory at ETH. Every sample consists of three identical

130 replicates. Filters were frozen after filtering and kept frozen until freeze-drying. Dried filters were re-weighed to obtain suspended sediment concentrations (SSC). Filtered water was collected for DOC in 120 mL pre-combusted (450°C, 6 h) amber bottles, acidified to pH 2 with 85 % H₃PO₄ (120 µL), and stored cooled and in the dark. About 4 mL of filtrate was collected in glass vials for the analyses of water isotopic compositions.

2.3 Geochemical analyses

- 135 Filter pieces (3 mm diameter) containing on average 350 μgC were placed in Ag boats (Säntis Analytical AG) and decarbonated in a desiccator under HCl vapor (70° C, 72 h), followed by neutralization with NaOH (70° C, 72 h; Freymond et al., 2018). Vapor-acid treated samples were wrapped in tin boats and analyzed for OC content and stable isotopic composition (δ¹³C) on an elemental analyzer-isotope ratio mass spectrometer (EA-IRMS, Elementar, Vario MICRO cube – Isoprime, VISION) at the Laboratory of Ion Beam Physics (LIP) at ETH Zurich. Radiocarbon (¹⁴C) was measured directly as
- 140 CO₂ gas using a mini radiocarbon dating system (MICADAS, Ionplus; Wacker et al., 2010; McIntyre et al., 2016). Samples were calibrated against Oxalic Acid II (NIST SRM 4990C) as well as an in-house soil and shale standards to correct for contamination during fumigation. All samples were corrected for constant contamination (~8 µgC) following Haghipour et al. (2019) and ¹⁴C data are reported as fraction modern, F¹⁴C (Reimer et al., 2004).

A wet chemical oxidation approach was used to convert DOC into CO₂ (Lang et al., 2012, 2016). In short, aliquots of DOC
(10-134 μgC) were oxidized using an acidified sodium persulfate solution (100 mL H₂O + 4.0 g Na₂S₂O₈ + 200 μL of 85 % H₃PO₄) and were purged with high-purity helium gas (Grade 5.0, 99.9999% pure, for 10 min) removing ambient air and inorganic CO₂. The samples were heated to 100° C for 1 h to convert the DOC to CO₂. The resulting CO₂ was analyzed using a MICADAS equipped with a gas-accepting ion source (LIP, ETH Zurich). Blank assessment was based on repeated measurements of sucrose (Sigma, δ¹³C: -12.4 ‰ VPDB, F¹⁴C: 1.053±0.003) and phthalic acid (Sigma, δ¹³C: -33.6 ‰ VPDB, F¹⁴C<0.0025) standards. The evaluation of constant contamination amounted to ~2 µgC (Haghipour et al., 2019).

Analysis of water isotopic compositions (δ^2 H and δ^{18} O) was performed on a Picarro L2120-i cavity-ringdown spectrometer (Geological Institute, ETH). Standards comprised VSMOW2, GISP, and SLAP2 as well as three in-house reference waters. Each sample and reference material were injected seven times while discarding the first three injections to eliminate instrumental memory effects.





155 2.4 Bayesian isotope mixing model

MixSIAR is an open-source Bayesian tracer mixing model framework in the R computing environment that allows the estimation of fractional contributions of multiple sources to a mixture (Stock and Semmens, 2016; Stock et al., 2018). The model accounts for uncertainties in the source endmember compositions. Although Bayesian mixing models were originally intended to constrain animal diets in ecology, they are often applied to apportion relative contributions of OC sources in

- 160 rivers and lakes (Butman et al., 2015; Upadhayay et al., 2017; Repasch et al., 2021). We parameterized the mixing model using POC-δ¹³C, POC-F¹⁴C, and three potential endmember compositions assessing the influence of seasonal and hydrodynamic variations on the source apportionment of organic matter in the Sihl River catchment. Organic carbon sources comprised leaf litter, soil (0-40 cm), and bedrock material collected within the Sihl catchment (**Table 2**) (Smith et al., 2013; van der Voort et al., 2016; Gies et al., 2022). MixSIAR was run without any initial assumptions (uninformative prior), a
- 165 burn-in of 200,000 iterations, a thinning factor of 100, and a chain length of 300,000 for three parallel Markov chain Monte Carlo chains. Model convergence was evaluated using Geweke (Geweke, 1991) and Gelman-Rubin metrics (Gelman and Rubin, 1992).

2.5 Estimating fluvial loads using traditional sediment rating curves

Water discharge is a key parameter to discern particulate matter transport in rivers. Commonly, rating curves are used to calculate fluvial export where the sample collection is insufficient to define continuous concentration records. The relationship between concentration (*C*) data and discharge (*Q*) is fitted with a power law function (e.g., Walling, 1977; Cohn, 1995; Syvitski et al., 2000; Wheatcroft et al., 2010), $C=aQ^b\varepsilon$, where the exponent *a* and *b* are rating coefficients inferred from a least linear squares regression of logarithmically transformed data. These rating curves are prone to underestimate long-term sediment transport rates by 10 to 50% (Asselman, 2000; Cohn, 1995; Ferguson, 1986). We apply Duan's (1983)

175 nonparametric retransformation bias correction factor (ε) appropriate for non-normal error distributions to compensate for this underestimation. Non-linear least squares regressions often replace traditional approaches and allow an unbiased estimation of rating curve parameters (Asselman, 2000).

2.6 Estimating fluvial loads using machine learning

Although machine learning techniques gain increasing popularity in environmental and earth sciences (e.g., Karpatne et al., 2019; Reichstein et al., 2019), their application in fluvial hydrology currently remains limited (Olyaie et al., 2015; Choubin et al., 2018; Sharafati et al., 2020). Machine (supervised) learning refers to a set of data mining approaches that develop pattern recognition based on a sample dataset in order to predict unlabeled target values. Here, we applied four commonly used machine learning algorithms: a multiple linear regression (MLR), a support vector regression machine (SVR) (Drucker et al., 1997), a random forest regression (RFR) (Breiman, 2001), and a neural network regression (NNR) (McCulloch and





185 Pitts, 1943). A detailed description of the applied models can be found in **Appendix B**. We evaluated and compared different techniques with the goal of estimating annual sediment and POC fluxes for the Sihl River basin.

A suite of predictor variables was used to construct regression models delineating sediment and OC export rates in the Sihl River. Variables include river discharge (Q), stage (H), precipitation (P), 1-day (P_{t-1}), and 2-day (P_{t-2}) antecedent precipitation spanning from 1974 to 2020. Daily discharge and water level values were obtained from the gauging station Sihlhölzli, Zurich,

- 190 operated by the Swiss Federal Office for the Environment (FOEN, https://www.hydrodaten.admin.ch). Daily precipitation data was retrieved for 21 stations located within and around the Sihl River basin from the Federal Office of Meteorology and Climatology (MeteoSwiss, https://gate.meteoswiss.ch). We applied inverse distance weighted (IDW) interpolation to produce inclusive and comprehensive maps describing the distribution of daily rainfall in the Sihl River watershed. The IDW approach, without considering orography, assumes that the attributed value of an unknown point is the weighted average of known values
- 195 within its neighborhood. Weights are inversely related to the distances between the predicted and sampled locations. Input variables a standardized using a robust scaler accounting for skewed data distributions and outliers.

All models were developed using scikit-learn, an open software machine learning library for the Python programming language. We tested several combinations of input predictors (**Table B1**) for machine learning approaches developing regression models. Hyperparameters for MLR, SVR, RFR, and NNR were determined using tuning techniques. Model tuning and stable model

- 200 results were derived by 10-fold nested cross-validation (trials=20). The performance of each model to predict suspended sediment and POC concentrations was evaluated based on three commonly used statistical metrics: the coefficient of determination (R²), the root mean squared error (RMSE), and the mean absolute error (MAE). While the R² indicates the precision of the standard regression type, RMSE and MAE represent the model accuracy. All models were visually examined and compared using a combination of violin and strip plots illustrating the probability and actual distributions of the observed
- 205 and predicted data. The best-performing algorithms were chosen to interpolate annual sediment and OC export rates. Predicted sediment and POC concentration values of <0 were set to 0.</p>

2.7 Statistical analyses

In order to statistically assess seasonal or rainfall-driven changes in exported sediment and OC concentrations and compositions, we introduce meteorological seasons and the discrimination between baseflow and stormflow conditions as

210 categorical variables. A discharge threshold value of 12.7 m³ s⁻¹ was derived from the average daily flow duration curve of the Sihl River spanning 47 yr of continuous observation (Fig. C1). A flow duration curve represents the frequency of occurrence of various flow rates. Recorded discharges are ranked according to their magnitude and subdivided into the percentages of time during which specific flows are equaled or exceeded. Flow rates ranging from 0 to 10% exceedance are categorized as high flow events, while values above 90% indicate the contributions of groundwater to the streamflow.





215 Due to the assumption violations of normality, equal variances, and equal sample sizes, we performed non-parametric Mann-Whitney and Kruskal-Wallis rank-sum tests. After the identification of significant between-group differences, we applied Conover-Iman post hoc tests with a Bonferroni adjustment of *p*-values (**Table D1**). All statistical comparisons are reported at the 95 % confidence interval (p<0.05).

3 Results

240

220 **3.1 Basin hydrology**

Mean annual discharges observed during the study period (2016-2019) are comparable to the long-term mean value of Q_{mean} 6.8±0.1 m³ s⁻¹ (M±SE). The lowest annual mean discharge is observed in 2018 (6.4±0.4 m³ s⁻¹), reflecting prolonged periods of drought (Hari et al., 2020; Peters et al., 2020). The highest annual mean discharge amounts to 7.1±0.4 m³ s⁻¹ in 2016. The sampled discharges range from 2.7 to 77 m³ s⁻¹ and represent the full range of discharge conditions observed during the 40-

225 month study period (Fig. 2a; Fig. C1). We observe no pronounced seasonal variability in the discharge of the Sihl River. Slight increases in water export coincide with snowmelt and periods of frequent storms in spring and summer (57%). The majority of the annual discharge occurs during storm events (Q>12.7 m³ s⁻¹; 82%), while baseflow conditions account for only 18%.

Riverine water isotopic compositions vary from -89.4 to -51.3 % for δ^2 H values and from -12.7 to -7.7 % for δ^{18} O values (Fig.

230 2c-d, 3a; Table S1). We note no difference between waters delivered during baseflow and stormflow conditions. However, water isotopic compositions are subject to seasonal shifts. While the majority of precipitation was primarily sourced from the North Atlantic, enriched H and O compositions indicate enhanced moisture supply from terrestrial Mediterranean and locally recycled moisture sources during the summer months (LeGrande and Schmidt, 2006; Batibeniz et al., 2020).

3.2 Suspended sediment and organic carbon concentrations

- Suspended sediment concentrations range between 0.8 to 133.1 mg L⁻¹, with an average of 13.5±2.8 mg L⁻¹ (n=77) during low flow conditions (Fig. 3b; Table S1). The export of SSC reached an observed maximum of 398.3 mg L⁻¹ (241.3±28.3 mg L⁻¹, n=17) during high discharge events. In comparison, SSC values of the Erlenbach varied from 19.8 to 15,310.7 mg L⁻¹ during stormflow (Smith et al., 2013). We observe higher sediment input rates in fall (61.8±21.6 mg L⁻¹, n=29) and winter (64.8±25.8mg L⁻¹, n=25) compared to the spring (42.1±21.9 mg L⁻¹, n=17) and summer (44.1±11.6 mg L⁻¹, n=23) months.
 - Particulate OC concentrations range from 0.01 to 12.08 mgC L⁻¹, with an average concentration of 1.37 ± 0.27 mgC L⁻¹ (n=90, **Table S1**). Substantially more organic matter is exported during storm-driven events (5.52 ± 0.81 mgC L⁻¹, n=17) than during baseflow conditions (0.41 ± 0.10 mgC L⁻¹, n=73). The observed range of suspended sediment OC contents vary from 0.37 to

Suspended sediments collected from Lake Sihl and the Alp River show concentrations ranging from 1.2 to 82.2 mg L⁻¹.



(n=3).



11.64 wt% (2.48±0.18 wt%, n=92; Fig. 2e, 3c). The mean POC content for low discharges amounts to 2.36±0.17 wt% (n=75),
while POC content rapidly increases (3.02±0.63 wt%, n=17) during storm-driven events. In contrast to SSC, measured OC contents are lower during the fall (2.33±0.27 wt%, n=29) and winter (2.04±0.18 wt%, n=25) months, and increase to 2.84±0.75 wt% (n =15) in spring and 2.91±0.36 wt% in summer (n=23). The Sihl River transports higher POC contents compared to values from the Erlenbach reported by Smith et al. (2013; 1.45±0.06 wt%, n=122) and Gies et al. (2022; 1.79±0.34 wt%, n=24). In contrast, POC contents in the Lümpenbach and Vogelbach of ~5.35 wt% exceed those of the Sihl River (Gies et al., 2022).
Observed POC concentrations and contents vary from 0.04 to 0.76 mg L⁻¹ and from 0.89 to 2.13 wt% in Lake Sihl (n=15). Measured organic matter concentrations in the Alp River average to 0.10±0.07 mg L⁻¹ and OC contents to 2.58±0.24 wt%

3.3 Isotopic composition of particulate and dissolved organic carbon

Sihl River POC-δ¹³C signatures across the time-series range from -30.1 to -25.8 ‰ averaging -27.7±0.1 ‰ (n=92; Fig. 2f;
3d; Table S1). No statistically significant differences between POC-δ¹³C and discharge are observed, but we note pronounced seasonality (Table D1). Higher POC-δ¹³C values are recorded during the summer (-27.1±0.1 ‰, n=23), whereas OC measured in spring exhibits on average the lowest δ¹³C values (-28.1±0.2 ‰, n=17). In contrast, POC-δ¹³C values in Lake Sihl (-29.7±0.5 ‰, n=15) and the Alp River (-28.0±0.3 ‰, n=2) are generally lower. Bulk POC-δ¹³C in the Sihl River overlaps with reported C₃ vegetation and soil biomass constituents (Kohn, 2010; Smith et al., 2013; Gies et al., 2022).

- Sihl River POC-F¹⁴C values range from 0.56 to 1.00 (0.87±0.01, n=91; Fig. 2g; Table S1), and display a statistically significant positive correlation with discharge (rs=0.43, p<0.001). Mean F¹⁴C values of 0.86±0.01 (n=75) are measured during low flow and increase to 0.91±0.01 (n=16; Fig. 3e) during high flow conditions, indicating storm-driven mobilization and entrainment of undegraded, biospheric POC to the Sihl River, the latter having been observed in tectonically active regimes (Lyons et al., 2002; Carey et al., 2005; Hilton et al., 2008a, 2010; Gomez et al., 2010). In contrast, suspended sediment POC-F¹⁴C values in
- 265 the mountainous Erlenbach average 0.65±0.08 (n=6) (Smith et al., 2013). Gies et al. (2022) report POC-F¹⁴C signatures for the Erlenbach, Lümpenbach, and Vogelbach of 0.64±0.22 (n=24), 0.80±0.17 (n=26), and 0.76±0.25 (n=27), respectively. The depletion in ¹⁴C values in the Erlenbach likely indicates substantial contributions of petrogenic OC. Nevertheless, the overall OC-F¹⁴C values of Sihl River POC are in good agreement with forested, temperate catchments characterized by minor inputs of organic-rich sedimentary bedrock and the absence of intense agricultural land use (Raymond et al., 2004; Longworth et al.,
- 2007; Goñi et al., 2013). The Alp River and Lake Sihl display POC-F¹⁴C values of 0.84±0.01 (n=3) and 0.79±0.02 (n=15). The DOC-F¹⁴C values vary from 0.52 to 1.16, with a mean of 0.95±0.01 (n=77; Fig. 2h, 3f; Table S1). Moderately aged DOC is observed in the summer months (0.90±0.03, n=17), whereas DOC enriched in ¹⁴C is discharged during fall (0.97±0.01, n=23). Similar to the F¹⁴C signature of POC, high precipitation events supply more modern DOC to the Sihl River (0.98±0.01, n=17). On average, Lake Sihl (0.97±0.01, n=15) and the Alp River (0.97±0.01, n=2) display slightly higher DOC-F¹⁴C
- 275 signatures than the Sihl River.





3.4 Performance of predictive models

All three machine learning algorithms outperform traditional rating curve models (**Fig. 4**; **Fig. B1-2**) in predicting suspended sediment and POC concentrations. The statistical performance of the evaluated models according to different scenarios is listed in **Table B1**. Traditional rating curves overestimate low values of suspended sediment and POC leading to poor

- 280 performance. Based on model performance criteria, SSC in the Sihl River depends primarily on discharge, water stage, and 1-day antecedent precipitation as predictor variables. Similar, scenarios that include discharge, water stage, precipitation, and 1-day antecedent precipitation appear to reliably reproduce measured POC concentrations. While discharge and water stage display the highest predictive power for instantaneous SSC, POC concentrations are more accurately described by water stage and 1-day antecedent precipitation. Random forest regression achieves the overall best fit with observed SSC (scenario
- 7; R²=0.85, RMSE=39.0), followed by SVR (scenario 2; R²=0.81, RMSE=43.8), MLR (scenario 8; R²=0.80, RMSE=45.6), and NNR (scenario 2; R²=0.75, RMSE=48.2). The highest coefficient of determination (R²=0.73) and the lowest root mean squared error (RMSE=1.2) for predicting POC concentrations are obtained from SVR (scenario 4). The performance of NNR (scenario 10; R²=0.70, RMSE=1.4), RFR (scenario 8; R²=0.68, RMSE=1.3), and MLR (scenario 8; R²=0.59, RMSE=1.5) captured observed POC variations with less accuracy.

290 **3.5** Annual fluxes and yields of suspended sediment and organic carbon

We calculate suspended sediment and POC export fluxes using continuous 47-year daily water discharge, stage, and precipitation records (see **Sect. 2.6**). Given our intermittent sampling design, we are not able to correct export fluxes for hysteresis effects or supply limitations (Wymore et al., 2019). We regard estimated sediment and POC budgets as conservative estimates, constraining a lower boundary.

- 295 Annual suspended sediment flux estimation for the Sihl River range from 17,789.8±1,041.5 (non-linear least squares power law) to 25,788.2±3,775.7 t yr⁻¹ (bias-corrected power law) (**Table B2**). Fluxes provided by the best fitting model (RFR) average at 25,166.5±1,055.8 t yr⁻¹. The majority of sediment export occurs during storm-driven events (72.9-93.0 %). We observe elevated export fluxes during summer (36.6-48.5 %) and spring (26.4-31.4 %) months corresponding to snowmelt and convective rainfall (Schmidt et al., 2019). The suspended sediment load in fall and winter varies between 12.0 and 18.4 %.
- 300 Similar to the export of suspended sediment, modeled annual POC fluxes assume values between 426.3±21.4 (non-linear least squares power law) and 762.7±121.1 t yr⁻¹ (bias-corrected power law) (**Table B2**). The mean POC load inferred from the SVR model amounts to 573.5±24.6 t yr⁻¹. Particulate organic carbon is primarily mobilized and transported downstream during high discharge events (66.0-94.9 %). The highest POC loads are transported during summer (31.1-49.7 %) and spring (26.2-32.5 %), while lower fluxes are observed in fall and winter (11.5-18.3 %).
- 305 The reservoir Lake Sihl is considered a sediment trap, efficiently retaining particulate matter delivered from the upper Sihl River watershed. Therefore, mean annual yield calculations were restricted to the lower Sihl River basin, including the Alp





and Biber catchments. We estimate annual yields between 93.6 ± 5.5 and 135.7 ± 19.9 t km⁻² yr⁻¹ (RFR: 132.4 ± 5.6 t km⁻² yr⁻¹) for suspended sediment and 2.2 ± 0.1 and 4.0 ± 0.6 t km⁻² yr⁻¹ (SVR: 3.0 ± 0.1 t km⁻² yr⁻¹) for POC.

4 Discussion

310 4.1 Organic carbon source contributions

Rivers integrate a mixture of POC comprising contemporary organic matter supplied by terrestrial and aquatic producers, aged soil-derived organic matter, and OC devoid of ¹⁴C released by weathered sedimentary bedrock (e.g., Hedges et al., 1986; Masiello and Druffel, 2001; Raymond et al., 2004; Blair and Aller, 2012). These sources have distinct carbon isotopic signatures and provide constraints on the contribution from different OC inputs. The Sihl River receives a uniform mixture

- 315 of fresh, aged, and ancient OC pools with relatively little variations as seasons progress (Fig. 5a). Biospheric carbon sources consist of allochthonous (e.g., vegetation, soils) and autochthonous (phytoplankton, benthic algae, aquatic macrophytes) inputs. Slightly more enriched POC-δ¹³C values in the summer months may indicate the contribution of freshwater C₃ plants (δ¹³C: ~-18 ‰; Chikaraishi, 2013). However, the coarse-grained riverbed substrate prevents the colonization of macrophytes resulting in poor aquatic vegetation in headwaters and allows only localized growth in the lower reaches of the Sihl River
- 320 (Känel et al., 2021). The formation of large-scale phytoplankton blooms and microbial biofilms is likely restricted by the low abundance of nutrients (Känel et al., 2021; Romaní et al., 2004; Battin, 1999) and limited light conditions in forested river segments (Boston and Hill, 1991). Algal growth is further disturbed by high discharge events resulting in river bed movement and the loss of algal mats (Schuwirth et al., 2008). From the above reasoning, we believe that instream biomass does not contribute significant amounts of OC to the Sihl River and would not bias our interpretations.
- 325 Soils can often be partitioned into several endmembers reflecting different stages of soil development as aging, microbial decomposition, and respiration introduce alterations to the isotopic composition of organic matter (Fernandez et al., 2003; Werth and Kuzyakov, 2010; Wang et al., 2015). However, Swiss shallow soils display relatively muted gradients in OC-¹⁴C signatures with increasing soil depth and between climatic regions (van der Voort et al., 2016). This relatively homogenous isotopic composition has been ascribed to the presence of bomb-derived OC in soil layers up to 40 cm depth (van der Voort et al.)
- 330 al., 2016, 2019). Van der Voort et al. (2019) suggest that percolation of dissolved organic carbon (as constrained via waterextractable OC measurements) may serve as an agent to propagate modern carbon to deeper soil layers in nonwaterlogged (aerobic) soils, resulting in a less pronounced age gradient with depth. This suggests that physical and chemical soil erosion processes deliver primarily modern dissolved and particulate organic matter to the Sihl River impeding the allocation of distinct sources. Consequently, for the purpose of this study, top (0-20 cm) and deeper (>20 cm) soil horizons are regarded as a single
- 335 source.



340



The legacy of bomb-¹⁴C is also evident in the DOC fractions retrieved from the Sihl River, which are consistently ¹⁴C-enriched relative to corresponding POC samples (**Fig. 5b**). Commonly, DOC is leached from vegetation and soils by precipitation and its residence time in fluvial systems is similar to that of water (Raymond and Bauer, 2001; Marwick et al., 2015). The overall modern OC-F¹⁴C signature implies that DOC is primarily sourced from throughfall and the assimilation with non-fossil OC stored in litter and shallow soil layers (Inamdar et al., 2011, 2012). A recent study by von Freyberg et al. (2018) investigated the outflow of Swiss catchments and found that the residence time of contributing groundwater is less than 2-3 months. Similar

- to the Alp, Biber, Erlenbach, and Vogelbach systems (von Freyberg et al., 2018), the Sihl River water isotopic compositions reflect seasonal cycles in precipitation and streamflow, implying that groundwater contributions are primarily sourced from recent rainfall events (**Fig. 2c-d**). The short residence time and the likely shallow flow paths result in limited fluid and solid
- 345 interactions impeding the dissolution and mobilization of moderately aged soil organic matter and favoring the export of percolating DOC derived from litter and organic-rich soil horizons.

Aged DOC is often associated with anthropogenic disturbances including deforestation, agriculture (Moore et al., 2013; Drake et al., 2019), atmospheric deposition (Stubbins et al., 2012; Spencer et al., 2014), and the release of petroleum and wastewater (Griffith et al., 2009; Regnier et al., 2013; Butman et al., 2015). Although we observe isolated ¹⁴C-depleted DOC

350 signals collected during the summer months, which could be ascribed to the localized introduction of petrogenic OC emanating from fertilizers, mineral oil, or sewage, the majority of the DOC isotopic compositions indicates an overall low degree of anthropogenic disturbances.

MixSIAR modeling results suggest that suspended sediments in the Sihl River are largely derived from biospheric sources (Fig. 6). Fresh plant-derived debris constitutes the primary POC input from winter to spring (48 to 50%), reflecting the

- 355 considerable extent of forests and grasslands in the Sihl River catchment (Fig. 1b). In the summer (68±6%; M±SD) and fall (58±8 %) months, the composition of suspended sediment is dominated by the contribution from eroding soils. This observation is in agreement with soil erosion risk modeling based on the Revised Universal Soil Loss Equation (RUSLE) (Schmidt et al., 2016, 2019). Soil loss peaks between July and September in response to high rainfall erosivity on Swiss grasslands (Schmidt et al., 2019, 2016). Extensive vegetation cover is insufficient to counteract water-driven erosion. Rock-
- 360 derived OC, abundantly supplied by the highly erodible such as the Eocene Wägital flysch in the Alp watershed, contributes about 12±1 % in fall and spring and 15±2 % in summer and winter. Similar source proportions have been observed in the headwaters of the Alp River (Gies et al., 2022).

4.2 Downstream evolution of particulate organic carbon

The isotopic composition of Lake Sihl (open symbols) in the upper watershed is distinctly different from the Sihl River (**Fig. 5a**). While suspended sediment at the inlet of Lake Sihl (Lake Sihl 1) resembles material from the lower Sihl River,

suspended sediments within the lake display more depleted ¹³C and ¹⁴C signatures. Depleted POC-¹³C signals in lakes can be attributed to enhanced aquatic productivity. The isotopic composition of planktonic freshwater algae can range from -40 to -





22 ‰ with the majority of reported δ^{13} C being <-28 ‰ (Chikaraishi, 2013 and references therein). Isotopic fractionation of phytoplankton biomass can be amplified in the presence of abundant dissolved inorganic carbon (DIC). Lake Sihl, a

- 370 moderately alkaline waterbody, receives DIC from weathering carbonaceous bedrocks (Allochthon nappes, Northern limestone Alps, Säntis zone, Fig. 1c) via surface runoff and groundwater inflow. This "hard water effect" further manifests itself in the decrease of POC-F¹⁴C signatures as the input of bedrock-derived, ¹⁴C-depleted DIC dilutes the carbon isotopic content of the water (Blattmann et al., 2019; Broecker and Walton, 1959; Keaveney and Reimer, 2012). Isotopic shifts in POC may also be caused by the selective uptake, decomposition, and preservation of organic matter (Lehmann et al., 2002,
- 375 2004). Kinetic isotope effects during enzymatic reactions lead to the enrichment or depletion of biomolecules relative to the bulk biomass (O'Leary, 1988). Carbohydrates and proteins often enriched in ¹³C are bioactive compounds and preferentially decomposed by microbes (Harvey et al., 1995; van Dongen et al., 2002). In contrast, lipids derived from plant tissue or phytoplankton exhibit in general more depleted δ^{13} C values and are more robust against degradation, leading to accumulation in the particulate fraction (Harvey et al., 1995; van Dongen et al., 2002).
- 380 Surface water samples collected from Lake Sihl 1 in summer and Lake Sihl 2 in winter and spring are characterized by high SSC, low OC contents, are enriched in ¹³C, and depleted in ¹⁴C (Fig. 5a). These signatures resemble those of the Erlenbach (Smith et al., 2013; Gies et al., 2022), suggesting enhanced contributions of petrogenic OC (e.g., Wägital flysch). These suspended sediment particles were likely entrained by the Sihl and Minster Rivers in response to storm events, forming sediment plumes in the epilimnion of Lake Sihl (Fig. 1). Fine-grained mineral soil- and bedrock-derived particles are advected
- 385 to the center of the lake, whereas coarser, waterlogged biospheric debris mobilized by surface runoff is likely deposited near the river inlets (Douglas et al., 2022).

In comparison to the Sihl River, headwater-sourced POC is highly variable and encompasses a large range of carbon isotopic compositions (**Fig. 5a**, **7**). Headwaters, in particular the Erlenbach, receive substantial contributions of petrogenic OC (up to \sim 40 % of total OC) and fall between modern C₃ plants and bedrock endmembers (Smith et al., 2013; Gies et al., 2022). In a

- 390 recent study, Hilton et al. (2021) used fluxes of dissolved Re, a redox-sensitive element, to constrain weathering intensities of petrogenic OC in the Erlenbach and Vogelbach basins. Findings suggest that ~40 % of OC contained in the Wägital flysch is lost to oxidative remineralization, allowing the majority of unweathered petrogenic POC to be eroded and entrained into adjacent streams. Despite the high supply of sediment and petrogenic OC, the signal of severely aged organic matter is gradually lost downstream. We attribute the gradual attenuation of headwater OC signals to (1) a declining input and
- 395 increasingly distal source of bedrock-derived sediments, (2) an enhanced contribution of modern biospheric OC, (3) abiotic, and (4) biotic processes modifying organic matter during transit.

Highly erodible Eocene flysch sequences are superseded by more competent Cretaceous flysch and molasse units, likely resulting in reduced erosion rates and a lower input of petrogenic OC. Simultaneously, the relative abundance of entrained litter and surface soils increases downstream, thereby diluting or replacing bedrock-derived particles (Feng et al., 2016;

400 Hemingway et al., 2017). Numerous physicochemical mechanisms dynamically influence the addition, removal, and exchange



405



of OC in dissolved and particulate pools. These processes include flocculation-deflocculation, particle sorption-desorption, aggregation-disaggregation, leaching, settling, and photo-oxidation (Bauer and Bianchi, 2012; Bianchi and Bauer, 2012). Flocculation and adsorption of largely modern DOC (**Fig. 5b**) onto particles may provide an additional source of biospheric organic matter further masking petrogenic OC inputs (von Wachenfeldt and Tranvik, 2008; Attermeyer et al., 2018). Although rock-derived carbon is regarded as inert, persisting in the environment for at least millennia, studies have shown that microbes in aquatic settings can assimilate and efficiently respire OC devoid of ¹⁴C to CO₂ (Petsch et al., 2001; McCallister et al., 2004;

- Bouchez et al., 2010). However, flume experiments demonstrate that in-river transport, particle abrasion, and turbulent mixing exert minimal controls on the loss of organic matter and that the preservation of OC is primarily regulated by transient storage in floodplains (Scheingross et al., 2019, 2021). Considering the absence of extensive floodplains and the short transit times in
- 410 the Sihl River catchment, we regard oxidative loss as a minor factor contributing to the removal of bedrock-sourced POC (Fox et al., 2020) and assume that overall OC fluxes experience little microbial decomposition during active fluvial transfer.

The downstream exchange and dilution of petrogenic OC with undegraded organic matter have previously been observed in large river systems such as the Amazon (Hedges et al., 1986, 2000; Mayorga et al., 2005), the Ganges-Brahmaputra (Galy et al., 2008; Galy and Eglinton, 2011), the Congo (Hemingway et al., 2017), and Orange rivers (Herrmann et al., 2016). The

415 alteration of riverine POC composition in these extensive river networks occurs over large spatial scales involving changes in topography, basin morphology, geology, vegetation, and climatic variables. In comparison, the Sihl River integrates and modifies exported POC over a ~40 km river interval without experiencing significant shifts in basin characteristics. These findings imply that low-order rivers may possess the potential to actively transform exported OC impacting local and regional terrestrial carbon cycles.

420 4.3 Hydrologic controls on particulate organic carbon sources and pathways

MixSIAR model results suggest that storm-driven events mobilize and flush enhanced proportions of plant-derived material into the Sihl River, with values rising from 32 ± 9 % during baseflow to 51 ± 12 % during high flow conditions. Concurrently, relative inputs of soil and petrogenic OC decrease from 54 ± 10 % to 39 ± 12 % and from 14 ± 1 % to 10 ± 1 %, respectively.

We observe pronounced patterns in the character of POC isotopic compositions as a function of discharge (Fig. 7). During low

- 425 flow, POC-δ¹³C and POC-F¹⁴C values display a large spread in values, corresponding to heterogeneous contributions from a variety of potential sources. In contrast, the isotopic signatures of storm-derived POC are less variable and appear to converge (as indicated by the arrows in Fig. 7b-c) to a POC-δ¹³C value of -27.5±0.1 ‰ and a POC-F¹⁴C value of 0.90±0.01. Similar behavior is noted in the Sihl headwaters. Although POC exhibits a larger variance in these headwaters, the composition of OC isotopes forms a relatively homogenous mixture during elevated precipitation events (Gies et al., 2022). This convergence
- 430 might indicate a thorough mixing of several carbon pools mobilized during a storm event (Kao and Liu, 2000; Hilton et al., 2008a; Gies et al., 2022). However, higher OC contents and predominately modern ¹⁴C signatures of the Sihl River suspended load point towards a marked shift in sources and transport pathways from moderately aged to fresher organic carbon pools





primarily consisting of surface soils and litter (Fig. 3 c-d, 6 a). The enhanced storm-facilitated export of modern biospheric material has been previously observed in subtropical (Hilton et al., 2008b, 2012; Wang et al., 2016; Qiao et al., 2020) and

- 435 temperate regions (Medeiros et al., 2012; Hatten et al., 2012; Goñi et al., 2013), and has been attributed to increased surface runoff and landsliding. Heavy rainfall and the resulting overland flow mobilize and laterally transport loose plant-derived debris and sediment from surface soils to adjacent fluvial systems (Harmon et al., 1986; Medeiros et al., 2012; Hatten et al., 2012; Turowski et al., 2013). Storm-induced erosion processes such as shallow landslides efficiently detach litter and organicrich surface soil layers and actively connect forested hillslopes to river channels (Hovius et al., 2000; Hilton et al., 2011b,
- 440 2012). Storm events may also alter the relative contributions of distal and proximal OC sources. Rising water levels inundate adjacent riparian zones, potentially mobilizing significant amounts of standing riparian biomass, litter, and soil organic matter (Marwick et al., 2014; Sutfin et al., 2016). However, as the vegetation in the Sihl River watershed consists primarily of C₃ plants, the distinction between proximal and distal biospheric sources cannot be resolved by a simple carbon isotopic approach.

Best-fit model parameters and predictor variables constraining suspended sediment and POC concentrations show marked

- differences indicating a divergence of sources and mobilization pathways as a function of discharge. The traditional rating curve exponent *b* is often interpreted as a proxy for the mobilization rate of particles in a fluvial system. The slightly higher rating curve exponent for POC (1.9 ± 0.1) compared to SSC (1.8 ± 0.1) indicates that the relative export of organic matter may exceed the export of suspended sediment during elevated discharges. Discharge is the primary descriptive variable predicting SSC in all machine learning techniques followed by water stage and 1-day antecedent precipitation (**Table B1**). In contrast,
- 450 the predictor variables water stage and 1-day antecedent precipitation achieve high performance for POC. Similar to the isotopic evidence, enhanced inputs of biospheric OC driven by high precipitation/flooding events may point to increased surface runoff and water erosivity entraining plant- and soil-derived material, while carbon-poor sediment deposited as bank or bedload is likely remobilized within the channel system.

4.4 Export fluxes and implications

- 455 Sediment fluxes in the Sihl River basin are less than half of the particle export documented in other Swiss Rivers (Spreafico et al., 2005). The low yield can be attributed to (1) the damming of the upper basin (Sihl Lake), retaining annually up to 1470 t yr⁻¹ of the suspended sediment load (Spreafico, 2007), (2) fortified banks, (3) topography, and (4) catchment geology. The lithology in the lower watershed consists mainly of molasse characterized by low slope angles and a reduced erosion potential (Schuerch et al., 2006; Korup and Schlunegger, 2009). Only the Alp River drains highly erosive flysch formations
- 460 in the lower Sihl River basin (Winkler et al., 1985). However, flysch units differ markedly in their erosivity. While the Erlenbach is underlain by easily erodible, fine-grained Eocene pelitic turbidites and mudstone sequences, bedrock in the Lümpenbach and Vogelbach watersheds mainly consist of more competent Cretaceous calcareous sandstones (Fig. 1d) (Keller and Weibel, 1991; Milzow et al., 2006; Schuerch et al., 2006). These differences in lithological units are manifested in their respective sediment yields. The Erlenbach, although comprising 0.4 % of the lower Sihl River basin, supplies about





- 465 4.7 to 6.7 % of the overall particulate load, with mean annual sediment yields ranging from 1225 to 1648 t km⁻² y⁻¹ (Keller and Weibel, 1991; Smith et al., 2013). In comparison, the Vogelbach, with a watershed size twice as large as the Erlenbach, displays lower annual sediment yields of 725 t km⁻² yr⁻¹ (Keller and Weibel, 1991) which roughly amounts to similar SSC export flux rates (4.4-6.4 %).
- By multiplying OC fluxes with the mean values of the MixSIAR posterior distributions for baseflow and stormflow conditions,
 while neglecting contributions from in-situ aquatic productivity, we can estimate export rates for the contributions of vegetation, soil, and bedrock. To further evaluate obtained fluxes and yields, we extract net primary productivity (NPP), soil erosion rates, and soil OC contents for the lower Sihl River basin.

Net primary productivity is defined as the uptake and incorporation of carbon from the atmosphere into plant biomass minus the loss to metabolism and maintenance (Clark et al., 2001). Annual, gap-filled NPP products at a 500 m pixel resolution

- 475 (MOD17A3HGFv006, MODIS/Terra; ORNL DAAC 2022) are retrieved and averaged over 20 years. The mean NPP for the lower Sihl River basin amounts to 748±11 tC km⁻² yr⁻¹ which is ~200 tC km⁻² yr⁻¹ higher than the NPP values derived from a Bayesian ecosystem model approach based on forest inventory and flux tower data of *P. abies* (540±150 tC km⁻² yr⁻¹) and *F. sylvatica* (530±100 tC km⁻² yr⁻¹) dominated forests (Trotsiuk et al., 2020). The discrepancy between those two estimates is observed globally and can be attributed to differences in methodology, the classification of the forest cover, and uncertainties
- 480 associated with heterogenous respiration, litter, and below-ground biomass (Park et al., 2021). Plant detritus exported as POC in the Sihl River accounts for only a fraction of the carbon sequestered and stored in vegetation (0.13 to 0.38 %) and is in good agreement with global estimates of fluvially exported terrestrial biospheric OC (Galy et al., 2015).

We extract and summarize monthly soil loss rates Schmidt et al. (2019) (82.8 ± 12.1 t km⁻² yr⁻¹), while average soil OC contents (0-120 cm, 1.7 ± 0.6 wt%) are derived from Nussbaum et al. (2014) and Solly et al. (2020). The resulting soil OC loss of 1.4 ± 0.5

- 485 tC km⁻² yr⁻¹ is consistent with the computed Sihl River soil export yield ranging from 1.0 to 1.6 tC km⁻² yr⁻¹. Similar soil OC erosion rates have been estimated for the Lümpenbach (0.7±0.3 tC km⁻² yr⁻¹) and Vogelbach (0.8±0.4 tC km⁻² yr⁻¹) watersheds (Gies et al., 2022). In contrast, the Erlenbach displays reduced soil OC yields of 0.3±0.2 tC km⁻² yr⁻¹. The geomorphology in the Erlenbach watershed rather favors the export of bedrock material due to abrasion and creep landslides (Schuerch et al., 2006; Smith et al., 2013).
- 490 Biospheric (2.0-3.6 tC km⁻² yr⁻¹) and petrogenic (0.3-0.4 tC km⁻² yr⁻¹) POC yields in the Sihl River are similar to contributions from the Erlenbach reported by Gies et al. (2022) (POC_{bio}: 1.2±0.4 tC km⁻² yr⁻¹) but distinctly lower than estimations by Smith et al. (2013) (POC_{bio}: 14.0±4.4 tC km⁻² yr⁻¹, POC_{petro}: 10.1±1.6 tC km⁻² yr⁻¹). Smith et al. (2013) focused their sample collection on storm-driven events yielding high export fluxes. However, the increase in relative biospheric contributions concomitant with the reduction in petrogenic OC proportions, and the decline in absolute export
- 495 yields from headwaters to the downstream Sihl sampling site, underline the impact of diverse processes acting upon and contributing to OC along the fluvial continuum. These processes reflect the impact of subtle changes in basin lithology





(erosivity, OC content), geomorphology (slope, floodplain extent), and anthropogenic activities (e.g., damming, channelization, land-use) on the age and the composition of exported POC.

5 Conclusions

- 500 This study focuses on temporal variations in organic carbon export from a Swiss subalpine river and provides insights into the mechanisms of sediment and associated OC mobilization and transport within a moderately steep river basin. Our results indicate that POC in the Sihl River consists primarily of modern to moderately aged biospheric OC derived from terrestrial vegetation and soils. In summer, severe storm events promote water erosivity and the entrainment of soil-sourced particles. While petrogenic carbon is prevalent in Sihl headwater catchments, the signal is gradually lost downstream. We associate this
- 505 decline of rock-derived OC with decreasing contributions of source material that are restricted to upstream segments of the watershed, the dilution and replacement by soil and plant biomass, and instream OC transformation processes. Despite the low stream gradient of the Sihl River, particle export is driven by episodic, short-lived storm events. We observe large variations in the isotopic composition of organic matter during baseflow conditions, whereas POC-δ¹³C and POC-F¹⁴C values converge to a more uniform mixture during storm-driven events. Results of traditional and machine learning modeling approach further
- 510 reveal diverging transport pathways for suspended sediment and OC with increasing discharge. Given the high POC content, the modern POC-F¹⁴C signature, and the differences in particle mobilization, we suggest that severe precipitation events facilitate the preferential entrainment of litter and surficial soil layers via surface runoff and shallow landsliding or via the increased input from proximal riparian vegetation and soils. Climate model simulations predict an increase in intensity, frequency, and duration of extreme precipitation events both regionally and globally with global warming (Myhre et al., 2019;
- 515 Kahraman et al., 2021) resulting in enhanced flood risks, water-induced erosion, and landsliding. The increased export of freshwater, nutrients, and sediment will likely severely affect downstream ecosystems, carbon cycling, requiring direct human intervention (Turowski et al., 2009; Talbot et al., 2018), and warranting the continuous monitoring of river systems.

Appendix A: Dissolved organic carbon concentrations

DOC concentration measurements were conducted using a Shimadzu system (TOC-L Series) at the Department of Environmental System Science at ETH Zurich and are reported in Table S1. However, measurements from August 2016 to March 2018 are not reported due to uncertainties in the quality of the measurements and we choose not to discuss them in the manuscript.





Appendix B: Machine learning approaches

Multiple linear regression (MLR) assumes a linear relationship between a single dependent continuous variable and several independent variables. To reduce overfitting of the MLR, we apply Elastic Net regularization which penalizes the model for both ℓ_1 and ℓ_2 -norms (Zou and Hastie, 2005).

Support vector regressions (SVR) are a popular machine learning approach performing linear or nonlinear classification, regression, and outlier detection. The support vector machine classification algorithm identifies an optimal hyperplane in ndimensional space to separate and categorize data points. In contrast, the SVR uses this principle to fit as many instances onto

530 a hyperplane while limiting margin violations. This supervised learning algorithm supports different kernels (linear, gaussian radial basis, polynomial) handling nonlinearity. Radial basis function kernels are commonly applied to fit non-linear regression lines and often outperform linear and polynomial kernels.

A standard decision tree is a non-parametric supervised learning method that predicts the value of a target variable by inferring simple decision rules based on data features. Decision trees are prone to overfitting the training set and thus are often replaced

535 by an ensemble of decision trees called a random forest. Random forests are generally built on bagging and random feature selection creating an uncorrelated forest of decision trees and thus generalizing well to unseen data (Breiman, 2001).

Neural Networks are a system of algorithms inspired by the human brain that attempts to recognize underlying patterns in a data set. The simplest neural network consists of an input and output layer that is interconnected through one hidden layer. Each neuron in these layers has an associated weight and threshold. The weighted sum of all neurons in a layer is passed

540 through an activation function and augmented by a bias term. In a feed-forward neural network, backpropagation adjusts the weights by minimizing the loss function and reducing the error between modeled and output values (Rumelhart et al., 1986). The utilized architecture consists of two hidden layers containing each ten neurons. To prevent overfitting, we apply dropout. This regularization technique temporarily removes units during the training period (Srivastava et al., 2014). Dropped units are chosen randomly.

545 Appendix C: Flow duration curve

Appendix D: Results of non-parametric analyses of variance

Data availability

All data generated are submitted and will be openly available in the EarthChem Library.





Supplement

550 Supporting information is available for this paper

Author contributions

MSS, ML, and TIE led the design of the study. HG and CF led data collection and analyses from 2014 to 2015. MSS conducted field and lab work from 2016 to 2019, data analysis and interpretation. NH contributed to laboratory analyses. MSS prepared the manuscript with contributions from all co-authors.

555 Competing interests

The authors declare that they have no conflict of interest.

Acknowledgments

We thank Francien Peterse for her motivation to initiate time-series sampling of the Sihl River. We thank Lena Märki for sample collection and Daniel Montluçon for laboratory assistance.

560 Financial support

M.S.S. was supported by the Swiss National Science Foundation through the grants SNF200020_163162/1, "CAPS-LOCK III" and SNF200020_184865/1, "CAPS-LOCK III".

References

Addor, N., Jaun, S., Fundel, F., and Zappa, M.: An operational hydrological ensemble prediction system for the city of Zurich
 (Switzerland): skill, case studies and scenarios, Hydrology and Earth System Sciences, 15, 2327–2347, https://doi.org/10.5194/hess-15-2327-2011, 2011.

Anon: Climate Change 2021 Working Group I contribution to the Sixth Assessment Report of the Intergovernmental Panel on Climate Change Summary for Policymakers, n.d.

Asselman, N. E. M.: Fitting and interpretation of sediment rating curves, Journal of Hydrology, 234, 228–248, https://doi.org/10.1016/S0022-1694(00)00253-5, 2000.

Attermeyer, K., Catalán, N., Einarsdottir, K., Freixa, A., Groeneveld, M., Hawkes, J. A., Bergquist, J., and Tranvik, L. J.: Organic Carbon Processing During Transport Through Boreal Inland Waters: Particles as Important Sites, Journal of Geophysical Research: Biogeosciences, 123, 2412–2428, https://doi.org/10.1029/2018JG004500, 2018.





 Aufdenkampe, A. K., Mayorga, E., Raymond, P. A., Melack, J. M., Doney, S. C., Alin, S. R., Aalto, R. E., and Yoo, K.: Riverine coupling of biogeochemical cycles between land, oceans, and atmosphere, Frontiers in Ecology and the Environment, 9, 53–60, https://doi.org/10.1890/100014.2011.

Batibeniz, F., Ashfaq, M., Önol, B., Turuncoglu, U. U., Mehmood, S., and Evans, K. J.: Identification of major moisture sources across the Mediterranean Basin, Climate Dynamics, 54, 4109–4127, https://doi.org/10.1007/s00382-020-05224-3, 2020.

Battin, T. J.: Hydrologic flow paths control dissolved organic carbon fluxes and metabolism in an alpine stream hyporheic zone, Water Resources Research, 35, 3159–3169, https://doi.org/10.1029/1999WR900144, 1999.

Battin, T. J., Luyssaert, S., Kaplan, L. A., Aufdenkampe, A. K., Richter, A., and Tranvik, L. J.: The boundless carbon cycle, Nature Geoscience, 2, 598–600, https://doi.org/10.1038/ngeo618, 2009.

Bauer, J. E. and Bianchi, T. S.: Dissolved Organic Carbon Cycling and Transformation, in: Treatise on Estuarine and Coastal Science, vol. 5, Elsevier Inc., 7–67, https://doi.org/10.1016/B978-0-12-374711-2.00502-7, 2012.

585 Bezzola, G. R. and Hegg, C.: Ereignisanalyse Hochwasser 2005, Teil 1 - Prozessse, Schäden und erste Einordnung, Umwelt Wissen, 0707, 215, https://doi.org/Umwelt-Wissen Nr. 0707, 2007.

Bianchi, T. S. and Bauer, J. E.: Particulate Organic Carbon Cycling and Transformation, in: Treatise on Estuarine and Coastal Science, vol. 5, Elsevier Inc., 69–117, https://doi.org/10.1016/B978-0-12-374711-2.00503-9, 2012.

Blair, N. E. and Aller, R. C.: The Fate of Terrestrial Organic Carbon in the Marine Environment, Annual Review of Marine Science, 4, 401–423, https://doi.org/10.1146/annurev-marine-120709-142717, 2012.

Blair, N. E., Leithold, E. L., Ford, S. T., Peeler, K. A., Holmes, J. C., and Perkey, D. W.: The persistence of memory: The fate of ancient sedimentary organic carbon in a modern sedimentary system, Geochimica et Cosmochimica Acta, 67, 63–73, https://doi.org/10.1016/S0016-7037(02)01043-8, 2003.

 Blattmann, T. M., Wang, S. L., Lupker, M., Märki, L., Haghipour, N., Wacker, L., Chung, L. H., Bernasconi, S. M., Plötze, M., and Eglinton, T. I.: Sulphuric acid-mediated weathering on Taiwan buffers geological atmospheric carbon sinks, Scientific Reports, 9, https://doi.org/10.1038/s41598-019-39272-5, 2019.

Boston, H. L. and Hill, W. R.: Photosynthesis-light relations of stream periphyton communities, Limnol. Oceanogr, 644-656 pp., 1991.

Bouchez, J., Beyssac, O., Galy, V. v., Gaillardet, J. J., France-Lanord, C., Maurice, L., Moreira, and Moreira-Turcq, P.: Oxidation of petrogenic organic carbon in the Amazon floodplain as a source of atmospheric CO2, Geology, 38, 255–258, https://doi.org/10.1130/G30608.1, 2010.

Breiman, L.: Random Forests, Machine Learning, 45, 5-32, https://doi.org/10.1007/978-3-030-62008-0 35, 2001.

Broecker, W. S. and Walton, A.: The Geochemistry of ¹⁴C in Freshwater Systems, Geochimica et Cosmochimica Acta, 16, 15–38, 1959.

Butman, D. E., Wilson, H. F., Barnes, R. T., Xenopoulos, M. A., and Raymond, P. A.: Increased mobilization of aged carbon to rivers by human disturbance, Nature Geoscience, 8, 112–116, https://doi.org/10.1038/ngeo2322, 2015.

605 Carey, A. E., Gardner, C. B., Goldsmith, S. T., Lyons, W. B., and Hicks, D. M.: Organic carbon yields from small, mountainous rivers, New Zealand, Geophysical Research Letters, 32, 1–5, https://doi.org/10.1029/2005GL023159, 2005.

Chikaraishi, Y.: ¹³C/¹²C Signatures in Plants and Algae, in: Treatise on Geochemistry: Second Edition, vol. 12, Elsevier Ltd., 95–123, https://doi.org/10.1016/B978-0-08-095975-7.01008-1, 2013.

Choubin, B., Darabi, H., Rahmati, O., Sajedi-Hosseini, F., and Kløve, B.: River suspended sediment modelling using the CART model: A comparative study of machine learning techniques, Science of the Total Environment, 615, 272–281, https://doi.org/10.1016/j.scitotenv.2017.09.293, 2018.

Clark, D. A., Brown, S., Kicklighter, D. W., Chambers, J. Q., Thomlinson, J. R., and Ni, J.: Measuring net primary production in forests: concepts and field methods, Ecological Applications, 356–370 pp., 2001.

Cohn, T. A.: Recent advances in statistical methods for the estimation of sediment and nutrient transport in rivers, Reviews of Geophysics, 33, 1117–1123, https://doi.org/10.1029/95RG00292, 1995.





Cole, J. J., Prairie, Y. T., Caraco, N. F., McDowell, W. H., Tranvik, L. J., Striegl, R. G., Duarte, C. M., Kortelainen, P., Downing, J. A., Middelburg, J. J., and Melack, J.: Plumbing the global carbon cycle: Integrating inland waters into the terrestrial carbon budget, Ecosystems, 10, 171–184, https://doi.org/10.1007/s10021-006-9013-8, 2007.

620 Coynel, A., Etcheber, H., Abril, G., Maneux, E., Dumas, J., and Hurtrez, J. E.: Contribution of small mountainous rivers to particulate organic carbon input in the Bay of Biscay, Biogeochemistry, 74, 151–171, https://doi.org/10.1007/s10533-004-3362-1, 2005.

van Dongen, B. E., Schouten, S., and Sinninghe Damsté, J. S.: Carbon isotope variability in monosaccharides and lipids of aquatic algae and terrestrial plants, Marine Ecology Progress Series, 232, 83–92, https://doi.org/10.3354/meps232083, 2002.

Doppler, T., Franssen, H. J. H., Kaiser, H. P., Kuhlman, U., and Stauffer, F.: Field evidence of a dynamic leakage coefficient for modelling river-aquifer interactions, Journal of Hydrology, 347, 177–187, https://doi.org/10.1016/j.jhydrol.2007.09.017, 2007.

625 Douglas, P. M. J., Stratigopoulos, E., Park, S., and Keenan, B.: Spatial differentiation of sediment organic matter isotopic composition and inferred sources in a temperate forest lake catchment, Chemical Geology, 603, https://doi.org/10.1016/j.chemgeo.2022.120887, 2022.

Drake, T. W., van Oost, K., Barthel, M., Bauters, M., Hoyt, A. M., Podgorski, D. C., Six, J., Boeckx, P., Trumbore, S. E., Cizungu Ntaboba, L., and Spencer, R. G. M.: Mobilization of aged and biolabile soil carbon by tropical deforestation, Nature Geoscience, 12, 541–546, https://doi.org/10.1038/s41561-019-0384-9, 2019.

630 Drucker, H., Surges, C. J. C., Kaufman, L., Smola, A., and Vapnik, V.: Support vector regression machines, in: Advances in Neural Information Processing Systems, 155–161, 1997.

Duan, N.: Smearing estimate: A nonparametric retransformation method, J Am Stat Assoc, 78, 605–610, https://doi.org/10.1080/01621459.1983.10478017, 1983.

Feng, X., Feakins, S. J., Liu, Z., Ponton, C., Wang, R. Z., Karkabi, E., Galy, V., Berelson, W. M., Nottingham, A. T., Meir, P., and West,
 A. J.: Source to sink: Evolution of lignin composition in the Madre de Dios River system with connection to the Amazon basin and offshore, Journal of Geophysical Research G: Biogeosciences, 121, 1316–1338, https://doi.org/10.1002/2016JG003323, 2016.

Ferguson, R. I.: River Loads Underestimated by Rating Curves, Water Resources Research, 22, 74–76, https://doi.org/10.1029/WR022i001p00074, 1986.

Fernandez, I., Mahieu, N., and Cadisch, G.: Carbon isotopic fractionation during decomposition of plant materials of different quality, Global Biogeochemical Cycles, 17, https://doi.org/10.1029/2001gb001834, 2003.

Fox, P. M., Bill, M., Heckman, K., Conrad, M., Anderson, C., Keiluweit, M., and Nico, P. S.: Shale as a Source of Organic Carbon in Floodplain Sediments of a Mountainous Watershed, Journal of Geophysical Research: Biogeosciences, 125, https://doi.org/10.1029/2019JG005419, 2020.

von Freyberg, J., Allen, S. T., Seeger, S., Weiler, M., and Kirchner, J. W.: Sensitivity of young water fractions to hydro-climatic forcing
 and landscape properties across 22 Swiss catchments, Hydrology and Earth System Sciences, 22, 3841–3861, https://doi.org/10.5194/hess-22-3841-2018, 2018.

Freymond, C. v., Lupker, M., Peterse, F., Haghipour, N., Wacker, L., Filip, F., Giosan, L., and Eglinton, T. I.: Constraining Instantaneous Fluxes and Integrated Compositions of Fluvially Discharged Organic Matter, Geochemistry, Geophysics, Geosystems, 19, 2453–2462, https://doi.org/10.1029/2018GC007539, 2018.

650 Galy, V. and Eglinton, T.: Protracted storage of biospheric carbon in the Ganges-Brahmaputra basin, Nature Geoscience, 4, 843–847, https://doi.org/10.1038/ngeo1293, 2011.

Galy, V., France-Lanord, C., and Lartiges, B.: Loading and fate of particulate organic carbon from the Himalaya to the Ganga-Brahmaputra delta, Geochimica et Cosmochimica Acta, 72, 1767–1787, https://doi.org/10.1016/j.gca.2008.01.027, 2008.

Galy, V., Peucker-Ehrenbrink, B., and Eglinton, T. I.: Global carbon export from the terrestrial biosphere controlled by erosion, Nature, 521, 204–207, https://doi.org/10.1038/nature14400, 2015.

Gelman, A. and Rubin, D. B.: Inference from Iterative Simulation Using Multiple Sequences, Statistical Science, 7, 457-472, 1992.

Geweke, J.: Evaluating the accuracy of sampling-based approaches to the calculation of posterior moments., No. 148. Federal Reserve Bank of Minneapolis., 30 pp., 1991.

Gies, H., Lupker, M., Wick, S., Haghipour, N., and Eglinton, T.: Discharge-modulated soil organic carbon export from 1 temperate mountainous headwater streams, Journal of Geophysical Research: Biogeosciences, https://doi.org/10.1002/essoar.10508072.1, 2022.



670

690

700



Golly, A., Turowski, J. M., Badoux, A., and Hovius, N.: Controls and feedbacks in the coupling of mountain channels and hillslopes, Geology, 45, 307–310, https://doi.org/10.1130/G38831.1, 2017.

Gomez, B., Baisden, W. T., and Rogers, K. M.: Variable composition of particle-bound organic carbon in steepland river systems, Journal of Geophysical Research: Earth Surface, 115, 1–9, https://doi.org/10.1029/2010JF001713, 2010.

665 Goñi, M. A., Hatten, J. A., Wheatcroft, R. A., and Borgeld, J. C.: Particulate organic matter export by two contrasting small mountainous rivers from the Pacific Northwest, U.S.A., Journal of Geophysical Research: Biogeosciences, 118, 112–134, https://doi.org/10.1002/jgrg.20024, 2013.

Goñi, M. A., Moore, E., Kurtz, A., Portier, E., Alleau, Y., and Merrell, D.: Organic matter compositions and loadings in soils and sediments along the Fly River, Papua New Guinea, Geochimica et Cosmochimica Acta, 140, 275–296, https://doi.org/10.1016/j.gca.2014.05.034, 2014.

Griffith, D. R., Barnes, R. T., and Raymond, P. A.: Inputs of fossil carbon from wastewater treatment plants to U.S. Rivers and oceans, Environmental Science and Technology, 43, 5647–5651, https://doi.org/10.1021/es9004043, 2009.

Grill, G., Lehner, B., Thieme, M., Geenen, B., Tickner, D., Antonelli, F., Babu, S., Borrelli, P., Cheng, L., Crochetiere, H., Ehalt Macedo, H., Filgueiras, R., Goichot, M., Higgins, J., Hogan, Z., Lip, B., McClain, M. E., Meng, J., Mulligan, M., Nilsson, C., Olden, J. D.,

675 Opperman, J. J., Petry, P., Reidy Liermann, C., Sáenz, L., Salinas-Rodríguez, S., Schelle, P., Schmitt, R. J. P., Snider, J., Tan, F., Tockner, K., Valdujo, P. H., van Soesbergen, A., and Zarfl, C.: Mapping the world's free-flowing rivers, Nature, 569, 215–221, https://doi.org/10.1038/s41586-019-1111-9, 2019.

Hagedorn, F., Bucher, J. B., and Schleppi, P.: Contrasting dynamics of dissolved inorganic and organic nitrogen in soil and surface waters of forested catchments with Gleysols, Geoderma, 100, 173–192, https://doi.org/10.1016/S0016-7061(00)00085-9, 2001.

680 Haghipour, N., Ausin, B., Usman, M. O., Ishikawa, N., Wacker, L., Welte, C., Ueda, K., and Eglinton, T. I.: Compound-Specific Radiocarbon Analysis by Elemental Analyzer-Accelerator Mass Spectrometry: Precision and Limitations, Analytical Chemistry, 91, 2042– 2049, https://doi.org/10.1021/acs.analchem.8b04491, 2019.

Hari, V., Rakovec, O., Markonis, Y., Hanel, M., and Kumar, R.: Increased future occurrences of the exceptional 2018–2019 Central European drought under global warming, Scientific Reports, 10, 1–10, https://doi.org/10.1038/s41598-020-68872-9, 2020.

685 Harmon, M. E., Franklin, J. F., Swanson, F. J., Sollins, P., Gregory, S. v., Lattin, J. D., Anderson, N. H., Cline, S. P., Aumen, N. G., Sedell, J. R., Lienkaemper, G. W., Cromack, K., and Cummins, K. W.: Ecology of Coarse Woody Debris in Temperate Ecosystems, Advances in Ecological Research, 15, 133–302, https://doi.org/10.1016/S0065-2504(03)34002-4, 1986.

Harvey, H. R., Tuttle, J. H., and Bell, J. T.: Kinetics of phytoplankton decay during simulated sedimentation: Changes in biochemical composition and microbial activity under oxic and anoxic conditions, Geochimica et Cosmochimica Acta, 59, 3367–3377, https://doi.org/10.1016/0016-7037(95)00217-N, 1995.

Hatten, J. A., Goñi, M. A., and Wheatcroft, R. A.: Chemical characteristics of particulate organic matter from a small, mountainous river system in the Oregon Coast Range, USA, Biogeochemistry, 107, 43–66, https://doi.org/10.1007/s10533-010-9529-z, 2012.

Hedges, J. I., Clark, W. A., Quay, P. D., Richey, J. E., Devol, A. H., and Santos, U. D. M.: Compositions and fluxes of particulate organic material in the Amazon River, Limnology and Oceanography, 31, 717–738, https://doi.org/10.4319/lo.1986.31.4.0717, 1986.

695 Hedges, J. I., Mayorga, E., Tsamakis, E., McClain, M. E., Aufdenkampe, A., Quay, P., Richey, J. E., Benner, R., Opsahl, S., Black, B., Pimentel, T., Quintanilla, J., and Maurice, L.: Organic matter in Bolivian tributaries of the Amazon River: A comparison to the lower mainstream, Limnology and Oceanography, 45, 1449–1466, https://doi.org/10.4319/lo.2000.45.7.1449, 2000.

Hemingway, J. D., Schefuß, E., Spencer, R. G. M., Dinga, B. J., Eglinton, T. I., McIntyre, C., and Galy, V. v.: Hydrologic controls on seasonal and inter-annual variability of Congo River particulate organic matter source and reservoir age, Chemical Geology, 466, 454–465, https://doi.org/10.1016/j.chemgeo.2017.06.034, 2017.

Herrmann, N., Boom, A., Carr, A. S., Chase, B. M., Granger, R., Hahn, A., Zabel, M., and Schefuß, E.: Sources, transport and deposition of terrestrial organic material: A case study from southwestern Africa, Quaternary Science Reviews, 149, 215–229, https://doi.org/10.1016/j.quascirev.2016.07.028, 2016.

Hilton, R. G., Galy, A., and Hovius, N.: Riverine particulate organic carbon from an active mountain belt: Importance of landslides, Global Biogeochemical Cycles, 22, 1–12, https://doi.org/10.1029/2006GB002905, 2008a.





Hilton, R. G., Galy, A., Hovius, N., Chen, M.-C., Horng, M.-J., and Chen, H.: Tropical-cyclone-driven erosion of the terrestrial biosphere from mountains, Nature Geoscience, 1, 759–762, https://doi.org/10.1038/ngeo333, 2008b.

Hilton, R. G., Galy, A., Hovius, N., Horng, M. J., and Chen, H.: The isotopic composition of particulate organic carbon in mountain rivers of Taiwan, Geochimica et Cosmochimica Acta, 74, 3164–3181, https://doi.org/10.1016/j.gca.2010.03.004, 2010.

710 Hilton, R. G., Galy, A., Hovius, N., Horng, M. J., and Chen, H.: Efficient transport of fossil organic carbon to the ocean by steep mountain rivers: An orogenic carbon sequestration mechanism, Geology, 39, 71–74, https://doi.org/10.1130/G31352.1, 2011a.

Hilton, R. G., Meunier, P., Hovius, N., Bellingham, P. J., and Galy, A.: Landslide impact on organic carbon cycling in a temperate montane forest, Earth Surface Processes and Landforms, 36, 1670–1679, https://doi.org/10.1002/esp.2191, 2011b.

Hilton, R. G., Galy, A., Hovius, N., Kao, S. J., Horng, M. J., and Chen, H.: Climatic and geomorphic controls on the erosion of terrestrial biomass from subtropical mountain forest, Global Biogeochemical Cycles, 26, 1–12, https://doi.org/10.1029/2012GB004314, 2012.

Hilton, R. G., Turowski, J. M., Winnick, M., Dellinger, M., Schleppi, P., Williams, K. H., Lawrence, C. R., Maher, K., West, M., and Hayton, A.: Concentration-Discharge Relationships of Dissolved Rhenium in Alpine Catchments Reveal Its Use as a Tracer of Oxidative Weathering, Water Resources Research, 57, https://doi.org/10.1029/2021WR029844, 2021.

Hovius, N., Stark, C. P., Hao-Tsu, C., and Jiun-Chuan, L.: Supply and removal of sediment in a landslide-dominated mountain belt: 720 Central Range, Taiwan, Journal of Geology, 108, 73–89, https://doi.org/10.1086/314387, 2000.

Inamdar, S., Singh, S., Dutta, S., Levia, D., Mitchell, M., Scott, D., Bais, H., and McHale, P.: Fluorescence characteristics and sources of dissolved organic matter for stream water during storm events in a forested mid-Atlantic watershed, Journal of Geophysical Research: Biogeosciences, 116, https://doi.org/10.1029/2011JG001735, 2011.

Inamdar, S., Finger, N., Singh, S., Mitchell, M., Levia, D., Bais, H., Scott, D., and McHale, P.: Dissolved organic matter (DOM)
 concentration and quality in a forested mid-Atlantic watershed, USA, Biogeochemistry, 108, 55–76, https://doi.org/10.1007/s10533-011-9572-4, 2012.

Jaun, S. and Ahrens, B.: Evaluation of a probabilistic hydrometeorological forecast system, Hydrology and Earth System Sciences Discussions, 6, 1843–1877, https://doi.org/10.5194/hessd-6-1843-2009, 2009.

Jochner, M., Turowski, J. M., Badoux, A., Stoffel, M., and Rickli, C.: The role of log jams and exceptional flood events in mobilizing coarse particulate organic matter in a steep headwater stream, Earth Surface Dynamics, 3, 311–320, https://doi.org/10.5194/esurf-3-311-2015, 2015.

Kahraman, A., Kendon, E. J., Chan, S. C., and Fowler, H. J.: Quasi-Stationary Intense Rainstorms Spread Across Europe Under Climate Change, Geophysical Research Letters, 48, https://doi.org/10.1029/2020GL092361, 2021.

Känel, B., Götz, C., Niederhauser, P., Sinniger, J., and Steinmann, P.: Zustand der Fliessgewässer von Limmat, Sihl und Zürichsee -Messkampagne 2020, 12 pp., 2021.

Kao, S. J. and Liu, K. K.: Stable carbon and nitrogen isotope systematics in a humandisturbed watershed (Lanyang-Hsi) in Taiwan and the estimation of biogenic particulate organic carbon and nitrogen fluxes, Global Biogeochemical Cycles, 14, 189–198, https://doi.org/10.1029/1999GB900079, 2000.

Karpatne, A., Ebert-Uphoff, I., Ravela, S., Babaie, H. A., and Kumar, V.: Machine Learning for the Geosciences: Challenges and
 Opportunities, IEEE Transactions on Knowledge and Data Engineering, 31, 1544–1554, https://doi.org/10.1109/TKDE.2018.2861006, 2019.

Keaveney, E. M. and Reimer, P. J.: Understanding the variability in freshwater radiocarbon reservoir offsets: A cautionary tale, Journal of Archaeological Science, 39, 1306–1316, https://doi.org/10.1016/j.jas.2011.12.025, 2012.

Keller, H. M. and Weibel, P.: Suspended Sediments in Streamwater - Indicators of Erosion and Bed Load Transport in Mountainous
 Basins. IAHS Publication No. 203, in: Sediment and Stream Water Quality in a Changing Environment: Trends and Explanation, 53–61, 1991.

Kohn, M. J.: Carbon isotope compositions of terrestrial C3 plants as indicators of (paleo)ecology and (paleo)climate., Proc Natl Acad Sci U S A, 107, 19691–5, https://doi.org/10.1073/pnas.1004933107, 2010.

Komada, T., Druffel, E. R. M., and Trumbore, S. E.: Oceanic export of relict carbon by small mountainous rivers, Geophysical Research Letters, 31, https://doi.org/10.1029/2004GL019512, 2004.



755



Korup, O. and Schlunegger, F.: Rock-type control on erosion-induced uplift, eastern Swiss Alps, Earth and Planetary Science Letters, 278, 278–285, https://doi.org/10.1016/j.epsl.2008.12.012, 2009.

Lang, S. Q., Bernasconi, S. M., and Früh-Green, G. L.: Stable isotope analysis of organic carbon in small (μg C) samples and dissolved organic matter using a GasBench preparation device, Rapid Communications in Mass Spectrometry, 26, 9–16, https://doi.org/10.1002/rcm.5287, 2012.

Lang, S. Q., Mcintyre, C. P., Bernasconi, S. M., Früh-Green, G. L., Voss, B. M., Eglinton, T. I., and Wacker, L.: Rapid ¹⁴C Analysis of Dissolved Organic Carbon in Non-Saline Waters, Radiocarbon, 58, 1–11, https://doi.org/10.1017/RDC.2016.17, 2016.

Lauerwald, R., Laruelle, G. G., Hartmann, J., Ciais, P., and Regnier, P. A. G.: Spatial patterns in CO₂ evasion from the global river network, Global Biogeochemical Cycles, 29, 534–554, https://doi.org/10.1002/2014GB004941, 2015.

760 Lauerwald, R., Regnier, P., Guenet, B., Friedlingstein, P., and Ciais, P.: How Simulations of the Land Carbon Sink Are Biased by Ignoring Fluvial Carbon Transfers: A Case Study for the Amazon Basin, One Earth, 3, 226–236, https://doi.org/10.1016/j.oneear.2020.07.009, 2020.

LeGrande, A. N. and Schmidt, G. A.: Global gridded data set of the oxygen isotopic composition in seawater, Geophysical Research Letters, 33, 1–5, https://doi.org/10.1029/2006GL026011, 2006.

765 Lehmann, M. F., Bernasconi, S. M., Barbieri, A., and Mckenzie, J. A.: Preservation of organic matter and alteration of its carbon and nitrogen isotope composition during simulated and in situ early sedimentary diagenesis, Geochimica et Cosmochimica Acta, 66, 3573– 3584, https://doi.org/10.1016/S0016-7037(02)00968-7, 2002.

Lehmann, M. F., Bernasconi, S. M., Barbieri, A., Simona, M., and McKenzie, J. A.: Interannual variation of the isotopic composition of sedimenting organic carbon and nitrogen in Lake Lugano: A long-term sediment trap study, Limnology and Oceanography, 49, 839–849, https://doi.org/10.4319/lo.2004.49.3.0839, 2004.

Leithold, E. L., Blair, N. E., and Perkey, D. W.: Geomorphologic controls on the age of particulate organic carbon from small mountainous and upland rivers, Global Biogeochemical Cycles, 20, 1–11, https://doi.org/10.1029/2005GB002677, 2006.

Leithold, E. L., Blair, N. E., and Wegmann, K. W.: Source-to-sink sedimentary systems and global carbon burial: A river runs through it, Earth-Science Reviews, 153, 30–42, https://doi.org/10.1016/j.earscirev.2015.10.011, 2016.

175 Li, M., Peng, C., Zhou, X., Yang, Y., Guo, Y., Shi, G., and Zhu, Q.: Modeling Global Riverine DOC Flux Dynamics From 1951 to 2015, Journal of Advances in Modeling Earth Systems, 11, 514–530, https://doi.org/10.1029/2018MS001363, 2019.

Longworth, B. E., Petsch, S. T., Raymond, P. A., and Bauer, J. E.: Linking lithology and land use to sources of dissolved and particulate organic matter in headwaters of a temperate, passive-margin river system, Geochimica et Cosmochimica Acta, 71, 4233–4250, https://doi.org/10.1016/j.gca.2007.06.056, 2007.

780 Lyons, W. B., Nezat, C. A., Carey, A. E., and Hicks, D. M.: Organic carbon fluxes to the ocean from high-standing islands, Geology, 30, 443–446, https://doi.org/10.1130/0091-7613(2002)030<0443:OCFTTO>2.0.CO;2, 2002.

Marwick, T. R., Borges, A. V., van Acker, K., Darchambeau, F., and Bouillon, S.: Disproportionate Contribution of Riparian Inputs to Organic Carbon Pools in Freshwater Systems, Ecosystems, 17, 974–989, https://doi.org/10.1007/s10021-014-9772-6, 2014.

Marwick, T. R., Tammoh, F., Teodoru, C. R., Borges, A. v., Darchambeau, F., and Bouillon, S.: The age of river-transported carbon: A global perspective, Global Biogeochemical Cycles, 29, 122–137, https://doi.org/10.1002/2014GB004911, 2015.

Masiello, A. and Druffel, E. R. M.: Carbon isotope geochemistry of the Santa Clara River, Global Biogeochemical Cycles, 15, 407–416, 2001.

 Mayorga, E., Aufdenkampe, A. K., Masiello, C. A., Krusche, A. v., Hedges, J. I., Quay, P. D., Richey, J. E., and Brown, T. A.: Young organic matter as a source of carbon dioxide outgassing from Amazonian rivers, Nature, 436, 538–541, https://doi.org/10.1038/nature03880, 2005.

McCallister, S. L., Bauer, J. E., Cherrier, J. E., and Ducklow, H. W.: Assessing sources and ages of organic matter supporting river and estuarine bacterial production: A multiple-isotope (Δ^{14} C, δ^{13} C, and δ^{15} N) approach, Limnology and Oceanography, 49, 1687–1702, https://doi.org/10.4319/lo.2004.49.5.1687, 2004.

McCulloch, W. S. and Pitts, W.: A logical calculus of the ideas immanent in nervous activity, Bulletin of Mathematical Biophysics, 5, 115–133, https://doi.org/10.1007/978-3-030-01370-7_61, 1943.





McIntyre, C. P., Lechleitner, F., Lang, S. Q., Haghiour, N., Fahrni, S., Wacker, L., and Synal, H. A.: ¹⁴C Contamination Testing in Natural Abundance Laboratories: A New Preparation Method Using Wet Chemical Oxidation and Some Experiences, Radiocarbon, 58, 935–941, https://doi.org/10.1017/RDC.2016.78, 2016.

Medeiros, P. M., Sikes, E. L., Thomas, B., and Freeman, K. H.: Flow discharge influences on input and transport of particulate and sedimentary organic carbon along a small temperate river, Geochimica et Cosmochimica Acta, 77, 317–334, https://doi.org/10.1016/j.gca.2011.11.020, 2012.

Milliman, J. D. and Farnsworth, K. L.: River Discharge to the Coastal Ocean. A global synthesis, 2nd ed., Cambridge University Press, Cambridge, New York, Melbourne, Madrid, Cape Town, Singapore, Sao Paulo, Dehli, Tokyo, Mexico City, 384 pp., 2013.

Milliman, J. D. and Syvitski, J. P. M.: Geomorphic/Tectonic Control of Sediment Discharge to the Ocean: The Importance of Small Mountainous Rivers, The Journal of Geology, 100, 525–544, https://doi.org/10.1086/629606, 1992.

Milzow, C., Molnar, P., McArdell, B. W., and Burlando, P.: Spatial organization in the step-pool structure of a steep mountain stream (Vogelbach, Switzerland), Water Resources Research, 42, https://doi.org/10.1029/2004WR003870, 2006.

Moore, S., Evans, C. D., Page, S. E., Garnett, M. H., Jones, T. G., Freeman, C., Hooijer, A., Wiltshire, A. J., Limin, S. H., and Gauci, V.: Deep instability of deforested tropical peatlands revealed by fluvial organic carbon fluxes, Nature, 493, 660–663, https://doi.org/10.1038/nature11818, 2013.

Myhre, G., Alterskjær, K., Stjern, C. W., Hodnebrog, Marelle, L., Samset, B. H., Sillmann, J., Schaller, N., Fischer, E., Schulz, M., and Stohl, A.: Frequency of extreme precipitation increases extensively with event rareness under global warming, Scientific Reports, 9, https://doi.org/10.1038/s41598-019-52277-4, 2019.

Nussbaum, M., Papritz, A., Baltensweiler, A., and Walthert, L.: Estimating soil organic carbon stocks of Swiss forest soils by robust external-drift kriging, Geoscientific Model Development, 7, 1197–1210, https://doi.org/10.5194/gmd-7-1197-2014, 2014.

O'Leary, M. H.: Carbon Isotopes in Photosynthesis, BioScience, 38, 328-336, https://doi.org/10.2307/1310735, 1988.

Olyaie, E., Banejad, H., Chau, K. W., and Melesse, A. M.: A comparison of various artificial intelligence approaches performance for estimating suspended sediment load of river systems: a case study in United States, Environmental Monitoring and Assessment, 187, https://doi.org/10.1007/s10661-015-4381-1, 2015.

820 Park, J. H., Gan, J., and Park, C.: Discrepancies between global forest net primary productivity estimates derived from modis and forest inventory data and underlying factors, Remote Sensing, 13, https://doi.org/10.3390/rs13081441, 2021.

Peters, W., Bastos, A., Ciais, P., and Vermeulen, A.: A historical, geographical and ecological perspective on the 2018 European summer drought, https://doi.org/10.1098/rstb.2019.0505, 26 October 2020.

Petsch, S. T., Eglinton, T. I., and Edwards, K. J.: ¹⁴C-dead living biomass: Evidence for microbial assimilation of ancient organic carbon during shale weathering, Science (1979), 292, 1127–1131, https://doi.org/10.1126/science.1058332, 2001.

Qiao, J., Bao, H., Huang, D., Li, D. W., Lee, T. Y., Huang, J. C., and Kao, S. J.: Runoff-driven export of terrigenous particulate organic matter from a small mountainous river: sources, fluxes and comparisons among different rivers, Biogeochemistry, 147, 71–86, https://doi.org/10.1007/s10533-019-00629-7, 2020.

Raymond, P. A. and Bauer, J. E.: Use of ¹⁴C and ¹³C natural abundances for evaluating riverine, estuarine, and coastal DOC and POC sources and cycling: A review and synthesis, Organic Geochemistry, 32, 469–485, https://doi.org/10.1016/S0146-6380(00)00190-X, 2001.

Raymond, P. A., Bauer, J. E., Caraco, N. F., Cole, J. J., Longworth, B., and Petsch, S. T.: Controls on the variability of organic matter and dissolved inorganic carbon ages in northeast US rivers, Marine Chemistry, 92, 353–366, https://doi.org/10.1016/j.marchem.2004.06.036, 2004.

Raymond, P. A., Hartmann, J., Lauerwald, R., Sobek, S., McDonald, C., Hoover, M., Butman, D., Striegl, R., Mayorga, E., Humborg, C.,
 Kortelainen, P., Dürr, H., Meybeck, M., Ciais, P., and Guth, P.: Global carbon dioxide emissions from inland waters, Nature, 503, 355–359, https://doi.org/10.1038/nature12760, 2013.

Regnier, P., Friedlingstein, P., Ciais, P., Mackenzie, F. T., Gruber, N., Janssens, I. a., Laruelle, G. G., Lauerwald, R., Luyssaert, S., Andersson, A. J., Arndt, S., Arnosti, C., Borges, A. v., Dale, A. W., Gallego-Sala, A., Goddéris, Y., Goossens, N., Hartmann, J., Heinze, C., Ilyina, T., Joos, F., LaRowe, D. E., Leifeld, J., Meysman, F. J. R., Munhoven, G., Raymond, P. a., Spahni, R., Suntharalingam, P., and

840 Thullner, M.: Anthropogenic perturbation of the carbon fluxes from land to ocean, Nature Geoscience, 6, 597–607, https://doi.org/10.1038/ngeo1830, 2013.



850

860



Regnier, P., Resplandy, L., Najjar, R. G., and Ciais, P.: The land-to-ocean loops of the global carbon cycle, https://doi.org/10.1038/s41586-021-04339-9, 17 March 2022.

Reichstein, M., Camps-Valls, G., Stevens, B., Jung, M., Denzler, J., Carvalhais, N., and Prabhat: Deep learning and process understanding 845 for data-driven Earth system science, Nature, 566, 195-204, https://doi.org/10.1038/s41586-019-0912-1, 2019.

Reimer, P. J., Brown, T. A., and Reimer, R. W.: Discussion: Reporting and Calibration of Post-Bomb ¹⁴C Data, Radiocarbon, 46, 1290-1304, https://doi.org/10.2458/azu js rc.46.4183, 2004.

Repasch, M., Scheingross, J. S., Hovius, N., Lupker, M., Wittmann, H., Haghipour, N., Gröcke, D. R., Orfeo, O., Eglinton, T. I., and Sachse, D.: Fluvial organic carbon cycling regulated by sediment transit time and mineral protection, Nature Geoscience, 14, 842-848, https://doi.org/10.1038/s41561-021-00845-7, 2021.

Rickenmann, D., Turowski, J. M., Fritschi, B., Klaiber, A., and Ludwig, A.: Bedload transport measurements at the Erlenbach stream with geophones and automated basket samplers, Earth Surface Processes and Landforms, 37, 1000-1011, https://doi.org/10.1002/esp.3225, 2012.

Romaní, A. M., Guasch, H., Muñoz, I., Ruana, J., Vilalta, E., Schwartz, T., Emtiazi, F., and Sabater, S.: Biofilm structure and function and 855 possible implications for riverine DOC dynamics, Microbial Ecology, 47, 316-328, https://doi.org/10.1007/s00248-003-2019-2, 2004.

Rowland, R., Inamdar, S., and Parr, T.: Evolution of particulate organic matter (POM) along a headwater drainage: role of sources, particle size class, and storm magnitude, Biogeochemistry, 133, 181–200, https://doi.org/10.1007/s10533-017-0325-x, 2017.

Ruiz-Villanueva, V., Mazzorana, B., Bladé, E., Bürkli, L., Iribarren-Anacona, P., Mao, L., Nakamura, F., Ravazzolo, D., Rickenmann, D., Sanz-Ramos, M., Stoffel, M., and Wohl, E.: Characterization of wood-laden flows in rivers, Earth Surface Processes and Landforms, 44, 1694-1709, https://doi.org/10.1002/esp.4603, 2019.

Rumelhart, D. E., Hinton, G. E., and Williams, R. J.: Learning representations by back-propagating errors, Nature, 323, 533-5.36, 1986.

Scheingross, J. S., Hovius, N., Dellinger, M., Hilton, R. G., Repasch, M., Sachse, D., Gröcke, D. R., Vieth-Hillebrand, A., and Turowski, J. M.: Preservation of organic carbon during active fluvial transport and particle abrasion, Geology, 47, 958-962, https://doi.org/10.1130/G46442.1, 2019.

865 Scheingross, J. S., Repasch, M. N., Hovius, N., Sachse, D., Lupker, M., Fuchs, M., Halevy, I., Gröcke, D. R., Golombek, N. Y., Haghipour, N., Eglinton, T. I., Orfeo, O., and Schleicher, A. M.: The fate of fluvially-deposited organic carbon during transient floodplain storage, Earth and Planetary Science Letters, 561, https://doi.org/10.1016/j.epsl.2021.116822, 2021.

Schleppi, P., Muller, N., Feven, H., Papritz, A., Bucher, J. B. J. B., Fliihler, H., and Flühler, H.: Nitrogen budgets of two small experimental forested catchments at Alptal, Switzerland, Forest Ecology and Management, 101, 177-185, https://doi.org/10.1016/S0378-

870 1127(97)00134-5, 1998.

Schleppi, P., Waldner, P. A., and Fritschi, B.: Accuracy and precision of different sampling strategies and flux integration methods for runoff water: Comparisons based on measurements of the electrical conductivity, Hydrological Processes, 20, 395–410, https://doi.org/10.1002/hyp.6057, 2006.

Schmidt, S., Alewell, C., Panagos, P., and Meusburger, K.: Regionalization of monthly rainfall erosivity patterns in Switzerland, 875 Hydrology and Earth System Sciences, 20, 4359–4373, https://doi.org/10.5194/hess-20-4359-2016, 2016.

Schmidt, S., Alewell, C., and Meusburger, K.: Monthly RUSLE soil erosion risk of Swiss grasslands, Journal of Maps, 15, 247-256, https://doi.org/10.1080/17445647.2019.1585980, 2019.

Schuerch, P., Densmore, A. L., McArdell, B. W., and Molnar, P.: The influence of landsliding on sediment supply and channel change in a steep mountain catchment, Geomorphology, 78, 222–235, https://doi.org/10.1016/j.geomorph.2006.01.025, 2006.

880 Schuwirth, N., Kühni, M., Schweizer, S., Uehlinger, U., and Reichert, P.: A mechanistic model of benthos community dynamics in the River Sihl, Switzerland, Freshwater Biology, 53, 1372–1392, https://doi.org/10.1111/j.1365-2427.2008.01970.x, 2008.

Schwab, M. S., Hilton, R. G., Haghipour, N., Baronas, J. J., and Eglinton, T. I.: Vegetal Undercurrents-Obscured Riverine Dynamics of Plant Debris, Journal of Geophysical Research: Biogeosciences, 127, https://doi.org/10.1029/2021jg006726, 2022.

Sharafati, A., Haji Seyed Asadollah, S. B., Motta, D., and Yaseen, Z. M.: Application of newly developed ensemble machine learning 885 models for daily suspended sediment load prediction and related uncertainty analysis, Hydrological Sciences Journal, 2022–2042, https://doi.org/10.1080/02626667.2020.1786571, 2020.



925



Smith, J. C., Galy, A., Hovius, N., Tye, A. M., Turowski, J. M., and Schleppi, P.: Runoff-driven export of particulate organic carbon from soil in temperate forested uplands, Earth and Planetary Science Letters, 365, 198–208, https://doi.org/10.1016/j.epsl.2013.01.027, 2013.

Solly, E. F., Weber, V., Zimmermann, S., Walthert, L., Hagedorn, F., and Schmidt, M. W. I.: A Critical Evaluation of the Relationship
 Between the Effective Cation Exchange Capacity and Soil Organic Carbon Content in Swiss Forest Soils, Frontiers in Forests and Global
 Change, 3, https://doi.org/10.3389/ffgc.2020.00098, 2020.

Spencer, R. G. M., Guo, W., Raymond, P. A., Dittmar, T., Hood, E., Fellman, J., and Stubbins, A.: Source and biolability of ancient dissolved organic matter in glacier and lake ecosystems on the Tibetan plateau, Geochimica et Cosmochimica Acta, 142, 64–74, https://doi.org/10.1016/j.gca.2014.08.006, 2014.

895 Spreafico, M.: Environmental impact caused by reservoir sedimentation management Advanced Training Workshop on Reservoir Sedimentation Management, 10-16 October 2007, Beijing China, 2007.

Spreafico, M., Lehmann, Ch., Jakob, A., and Grasso, A.: Feststoffbeobachtung in der Schweiz, 101 pp., 2005.

Srivastava, N., Hinton, G., Krizhevsky, A., and Salakhutdinov, R.: Dropout: A Simple Way to Prevent Neural Networks from Overfitting, Journal of Machine Learning Research, 1929–1958 pp., 2014.

900 Stock, B. C. and Semmens, B. X.: MixSIAR GUI User Manual. Version 3.1., https://doi.org/doi:10.5281/zenodo.1209993, 2016.

Stock, B. C., Jackson, A. L., Ward, E. J., Parnell, A. C., Phillips, D. L., and Semmens, B. X.: Analyzing mixing systems using a new generation of Bayesian tracer mixing models, PeerJ, 6:e5096, https://doi.org/10.7717/peerJ.5096, 2018.

Stubbins, A., Hood, E., Raymond, P. A., Aiken, G. R., Sleighter, R. L., Hernes, P. J., Butman, D., Hatcher, P. G., Striegl, R. G., Schuster, P., Abdulla, H. A. N., Vermilyea, A. W., Scott, D. T., and Spencer, R. G. M.: Anthropogenic aerosols as a source of ancient dissolved organic matter in glaciers, Nature Geoscience, 5, 198–201, https://doi.org/10.1038/ngeo1403, 2012.

- Sutfin, N. A., Wohl, E. E., and Dwire, K. A.: Banking carbon: A review of organic carbon storage and physical factors influencing retention in floodplains and riparian ecosystems, Earth Surface Processes and Landforms, 41, 38–60, https://doi.org/10.1002/esp.3857, 2016.
- 910 Syvitski, J. P., Morehead, M. D., Bahr, D. B., and Mulder, T.: Estimating fluvial sediment transport: The rating parameters, Water Resources Research, 36, 2747–2760, https://doi.org/10.1029/2000WR900133, 2000.

Talbot, C. J., Bennett, E. M., Cassell, K., Hanes, D. M., Minor, E. C., Paerl, H., Raymond, P. A., Vargas, R., Vidon, P. G., Wollheim, W., and Xenopoulos, M. A.: The impact of flooding on aquatic ecosystem services, Biogeochemistry, 141, 439–461, https://doi.org/10.1007/s10533-018-0449-7, 2018.

- 915 Tranvik, L. J., Downing, J. A., Cotner, J. B., Loiselle, S. A., Striegl, R. G., Ballatore, T. J., Dillon, P., Finlay, K., Fortino, K., Knoll, L. B., 915 Kortelainen, P. L., Kutser, T., Larsen, S., Laurion, I., Leech, D. M., Leigh McCallister, S., McKnight, D. M., Melack, J. M., Overholt, E., 917 Porter, J. A., Prairie, Y., Renwick, W. H., Roland, F., Sherman, B. S., Schindler, D. W., Sobek, S., Tremblay, A., Vanni, M. J., Verschoor, A. M., von Wachenfeldt, E., and Weyhenmeyer, G. A.: Lakes and reservoirs as regulators of carbon cycling and climate, Limnology and 918 Oceanography, 54, 2298–2314, https://doi.org/10.4319/lo.2009.54.6_part_2.2298, 2009.
- 920 Trotsiuk, V., Hartig, F., Cailleret, M., Babst, F., Forrester, D. I., Baltensweiler, A., Buchmann, N., Bugmann, H., Gessler, A., Gharun, M., 920 Minunno, F., Rigling, A., Rohner, B., Stillhard, J., Thürig, E., Waldner, P., Ferretti, M., Eugster, W., and Schaub, M.: Assessing the response of forest productivity to climate extremes in Switzerland using model–data fusion, Global Change Biology, 26, 2463–2476, https://doi.org/10.1111/gcb.15011, 2020.

Turowski, J. M., Yager, E. M., Badoux, A., Rickenmann, D., and Molnar, P.: The impact of exceptional events on erosion, bedload transport and channel stability in a step-pool channel, Earth Surface Processes and Landforms, 34, 613–628, https://doi.org/10.1002/esp, 2009.

Turowski, J. M., Badoux, A., and Rickenmann, D.: Start and end of bedload transport in gravel-bed streams, Geophysical Research Letters, 38, 1–5, https://doi.org/10.1029/2010GL046558, 2011.

Turowski, J. M., Badoux, A., Bunte, K., Rickli, C., Federspiel, N., and Jochner, M.: The mass distribution of coarse particulate organic matter exported from an Alpine headwater stream, Earth Surface Dynamics, 1, 1–11, https://doi.org/10.5194/esurf-1-1-2013, 2013.

930 Turowski, J. M., Hilton, R. G., and Sparkes, R.: Decadal carbon discharge by a mountain stream is dominated by coarse organic matter, Geology, 44, 27–30, https://doi.org/10.1130/G37192.1, 2016.



940

960



Upadhayay, H. R., Bodé, S., Griepentrog, M., Huygens, D., Bajracharya, R. M., Blake, W. H., Dercon, G., Mabit, L., Gibbs, M., Semmens, B. X., Stock, B. C., Cornelis, W., and Boeckx, P.: Methodological perspectives on the application of compound-specific stable isotope fingerprinting for sediment source apportionment, https://doi.org/10.1007/s11368-017-1706-4, 1 June 2017.

935 van der Voort, T. S., Hagedorn, F., McIntyre, C., Zell, C., Walthert, L., Schleppi, P., Feng, X., and Eglinton, T. I.: Variability in ¹⁴C contents of soil organic matter at the plot and regional scale across climatic and geologic gradients, Biogeosciences, 13, 3427–3439, https://doi.org/10.5194/bg-13-3427-2016, 2016.

van der Voort, T. S. van der, Mannu, U., Hagedorn, F., McIntyre, C. P., Walthert, L., Schleppi, P., Haghipour, N., and Eglinton, T. I.: Dynamics of deep soil carbon - insights from ¹⁴C time series across a climatic gradient, Biogeosciences, 16, 3233–3246, https://doi.org/10.5194/bg-16-3233-2019, 2019.

von Wachenfeldt, E. and Tranvik, L. J.: Sedimentation in boreal lakes - The role of flocculation of allochthonous dissolved organic matter in the water column, Ecosystems, 11, 803–814, https://doi.org/10.1007/s10021-008-9162-z, 2008.

Wacker, L., Bonani, G., Friedrich, M., Hajdas, I., Kromer, B., Nemec, M., Ruff', M., Suter, M., Synal, H., and Vockenhuber, C.: Micadas: Routine and High-Precision Radiocarbon Dating, Radiocarbon, 52, 252–262, https://doi.org/10.2458/azu_js_rc.52.3660, 2010.

945 Walling, D. E.: Assessing the accuracy of suspended sediment rating curves for a small basin, Water Resources Research, 13, 531–538, https://doi.org/10.1029/WR013i003p00531, 1977.

Wang, G., Jia, Y., and Li, W.: Effects of environmental and biotic factors on carbon isotopic fractionation during decomposition of soil organic matter, Scientific Reports, 5, https://doi.org/10.1038/srep11043, 2015.

Wang, J., Jin, Z., Hilton, R. G., Zhang, F., Li, G., Densmore, A. L., Gröcke, D. R., Xu, X., and Joshua West, A.: Earthquake-triggered increase in biospheric carbon export from a mountain belt, Geology, 44, 471–474, https://doi.org/10.1130/G37533.1, 2016.

Waser, L. T., Ginzler, C., and Rehush, N.: Wall-to-Wall tree type mapping from countrywide airborne remote sensing surveys, Remote Sensing, 9, https://doi.org/10.3390/rs9080766, 2017.

Werth, M. and Kuzyakov, Y.: 13C fractionation at the root-microorganisms-soil interface: A review and outlook for partitioning studies, https://doi.org/10.1016/j.soilbio.2010.04.009, September 2010.

955 West, A. J., Lin, C. W., Lin, T. C., Hilton, R. G., Liu, S. H., Chang, C. T., Lin, K. C., Galy, A., Sparkes, R. B., and Hovius, N.: Mobilization and transport of coarse woody debris to the oceans triggered by an extreme tropical storm, Limnology and Oceanography, 56, 77–85, https://doi.org/10.4319/lo.2011.56.1.0077, 2011.

Wheatcroft, R. A., Goñi, M. A., Hatten, J. A., Pasternack, G. B., and Warrick, J. A.: The role of effective discharge in the ocean delivery of particulate organic carbon by small, mountainous river systems, Limnology and Oceanography, 55, 161–171, https://doi.org/10.4319/lo.2010.55.1.0161, 2010.

Winkler, W., Wildi, W., van Stuijvenberg, J., and Caron, C.: Wägital-Flysch et autres flyschs jenniques en Suisse Centrale. Stratigraphie, sédimentologie et comparaisons, Eclogae Geologicae Helvetiae, 7, 1–22, 1985.

Wohl, E.: Bridging the gaps: An overview of wood across time and space in diverse rivers, Geomorphology, 279, 3–26, https://doi.org/10.1016/j.geomorph.2016.04.014, 2017.

965 Wohl, E. and Ogden, F. L.: Organic carbon export in the form of wood during an extreme tropical storm, Upper Rio Chagres, Panama, Earth Surface Processes and Landforms, 38, 1407–1416, https://doi.org/10.1002/esp.3389, 2013.

Wohl, E., Dwire, K., Sutfin, N., Polvi, L., and Bazan, R.: Mechanisms of carbon storage in mountainous headwater rivers, Nature Communications, 3, 1–8, https://doi.org/10.1038/ncomms2274, 2012.

Wymore, A. S., Leon, M. C., Shanley, J. B., and McDowell, W. H.: Hysteretic response of solutes and turbidity at the event scale across forested tropical montane watersheds, Frontiers in Earth Science, 7, https://doi.org/10.3389/feart.2019.00126, 2019.

Zou, H. and Hastie, T.: Regularization and variable selection via the elastic net, Journal of the Royal Statistical Society. Series B: Statistical Methodology, 67, 301–320, https://doi.org/10.1111/j.1467-9868.2005.00503.x, 2005.





975 **Table 1:** Summary of river and watershed characteristics for the Sihl River, Erlenbach, Lümpenbach, and Vogelbach. Information for the Sihl River headwater streams is provided by Smith et al. (2013) and von Freyberg et al. (2018).

		Erlenbach	Lümpenbach	Vogelbach	Sihl - Sihlhölzli	
Basin size at gauging station (km²)ª Average slope (°)ª Mean catchment elevation (m a.s.l.)ª		0.73 0.88 23.9 19.6 1359 1336		1.58 28.9 1335	346.02 19.5 1041	
Average discharge (m ³ s ⁻¹) ^a Basin geology ^b		0.04 Eocene flysch, Cretaceous	0.05 Cretaceous flysch	0.07 Cretaceous flysch	6.83 Subalpine molasse, flysch,	
		flysch	·		limestone, evaporites, sandstone	
Land-use ^c	Settlements (%)	0	0	0	9.0	
	Agriculture (%)	0	0	0	1.1	
	Meadows, pastures (%)	21.6	81	30	38.9	
	Forest (%)	59.5	19	70	43.4	
	Water bodies (%)	0	0	0	1.3	
	Unproductive areas (%)	18.9	0	0	6.1	

^a Federal Office for the Environment, https://www.bafu.admin.ch/bafu/de/home.html

^b Federal Office of Topography swisstopo, https://www.swisstopo.admin.ch/

° Federal Statistic Office, https://www.bfs.admin.ch/bfs/en/home.html

980

 Table 2: Organic carbon endmember compositions used in the MixSIAR Bayesian model.

	POC-δ ¹³ C (‰)				POC-F ¹⁴ C	2	Reference				
	n M SD			n M SD							
Bedrock	22	-25.71	0.84	est.	0.00	0.01	Smith et al. (2013), Gies et al. (2022)				
Soil	33	-27.20	0.76	50	1.00	0.10	Smith et al. (2013), van der Voort et al. (2016), Gies et al. (2022)				
Vegetation	21	-27.37	1.62	1	1.02	0.01	Smith et al. (2013), Gies et al. (2022)				
Vegetation	21	-27.37	1.62	1	1.02	0.01	Smith et al. (2013), Gies et al. (2022)				







Figure 1: Sihl River basin showing (a) altitude, (b) the distribution of slope angles (https://www.swisstopo.admin.ch), (c) different land-use types (https://www.bfs.admin.ch), and (d) the underlying geology (https://www.swisstopo.admin.ch, Winkler et al., 1985). The sampling location for the Sihl River time series is indicated as a yellow symbol; the locations of the Lake Sihl and the Alp River sampling sites are shown as grey symbols.







31





Figure 2: Hydrographs for the sampling periods from May 2014 to February 2015 and August 2016 to March 2019: (a) hourly discharge values (m³ s⁻¹; https://www.hydrodaten.admin.ch) and (b) daily precipitation values for the Sihl River basin (mm; https://gate.meteoswiss.ch). Gray dots represent individual sampling campaigns. Water isotopic compositions, (c) δ²H (‰) and (d) δ¹⁸O (‰), are shown alongside (e) particulate organic carbon (POC) contents (wt%), (f) POC-δ¹³C (‰,), (g) POC-F¹⁴C, and (h) dissolved organic carbon (DOC) F¹⁴C. Dots are scaled to discharge.











Figure 3: Combined violin and strip plots of (a) δ^{18} O (‰), (b) suspended sediment concentrations (SSC, mg L⁻¹), (c) particulate organic carbon (POC) content (wt%), (d) POC- δ^{13} C (‰), (e) POC- F^{14} C, and (f) dissolved organic carbon (DOC) F^{14} C faceted for seasons and discharge. Violin plots depict rotated kernel density plots. White vertical lines indicate median values. Significant between-group differences are denoted with brackets and *p*-values (**Table D1**).



Figure 4: Combined violin and strip plots comparing model performances in predicting (a) suspended sediment (SSC, mg L⁻¹) and (b) particulate organic carbon (POC, mg L⁻¹) concentrations. Models approaches include traditional and non-linear least squares power law functions as well as multilinear regression (MLR), support vector regression (SVR), random forest regression (RFR), and neural network regression (NNR). Violin plots depict rotated kernel density plots, with white horizontal lines indicating median values.







Figure 5: Relationship between (a) particulate organic carbon (POC) $F^{14}C$ and POC- $\delta^{13}C$ (‰) values and (b) between dissolved organic carbon (DOC) and POC- $F^{14}C$ values. Gray crosses indicate samples from the Alptal headwater streams: Erlenbach, Lümpenbach, and Vogelbach (Smith et al., 2013; Gies et al., 2022). Sources for plant, soil, and bedrock endmembers are listed in Table 2. Dots are color-coded for seasons and scaled to discharge.

010







Figure 6: Boxenplots depicting MixSIAR-derived posterior distributions of bedrock, soil, and terrestrial vegetation endmember contributions to the bulk POC load according to seasons and changes in discharge. The innermost box is drawn at the lower and upper quartiles. Incrementally narrower boxes represent the lower and upper octiles, hexadeciles, and so forth. Outliers are denoted with diamonds, while black lines depict medians.







Figure 7: Relationship between runoff (mm d⁻¹), (a) particulate organic carbon (POC) content (wt%), (b) POC- δ^{13} C (‰), and (c) POC- F^{14} C. Gray crosses indicate samples from the Erlenbach, Lümpenbach, and Vogelbach (Gies et al., 2022; Smith et al., 2013). Circles are color-coded for seasons.





020 Table B1: Model performance in predicting suspended sediment and particulate organic carbon concentrations for the investigated period.

Model	Scenario	Model structure	R ²	RMSE	MAE	R ²	RMSE	MAE
				mg L ⁻¹	mg L ⁻¹		mg L ⁻¹	mg L ⁻¹
Suspended S	Sediment Cond	centration						
Power Law (b	Power Law (biased corrected)					Power Law (non-linear least squares)		
			0.628	130.027	43.355	0.777	49.534	33.086
Multiple Linear Regression (Elastic Net)						Random Forest Regression		
	1 2 3 4 5 6 7 8 9 10 11	$\begin{array}{l} SSC \sim O \\ SSC \sim O + H \\ SSC \sim Q + H + P \\ SSC \sim Q + H + P + P_{b1} \\ SSC \sim Q + H + P + P_{b1} + P_{b2} \\ SSC \sim P_{b1} \\ SSC \sim P_{b1} + Q \\ SSC \sim P_{b1} + Q \\ SSC \sim P_{b1} + Q + H \\ SSC \sim H \\ SSC \sim H \\ SSC \sim H \\ SSC \sim P \end{array}$	0.769 0.781 0.775 0.791 0.530 0.792 0.796 0.673 0.666 0.335	48.241 47.119 47.877 46.165 66.768 45.974 45.566 57.444 57.614 81.021	26.720 26.052 26.440 26.236 26.597 40.194 25.824 26.007 34.489 35.160 56.256	0.831 0.789 0.809 0.829 0.823 0.589 0.847 0.783 0.670 0.659 0.115	41.617 47.633 46.130 41.131 44.162 67.226 39.027 41.032 56.707 54.453 99.313	20.951 24.752 23.236 20.961 21.384 28.772 19.136 20.784 32.647 27.061 58.731
Support Vecto	or Regression					Neural Network Regression		
	1 2 3 4 5 6 7 8 9 10 11	$\begin{array}{l} SSC \sim Q \\ SSC \sim Q + H \\ SSC \sim Q + H + P \\ SSC \sim Q + H + P + P_{k1} \\ SSC \sim Q + H + P + P_{k1} + P_{k2} \\ SSC \sim Q_{k1} \\ SSC \sim P_{k1} + Q \\ SSC \sim P_{k1} + Q \\ SSC \sim P_{k1} + Q + H \\ SSC \sim H \\ SSC \sim H + P_{k1} \\ SSC \sim P_{k2} \\ SSC \sim P_{k1} \\ SSC \sim P_{k2} \\ SSC \sim P_{k1} \\$	0.796 0.810 0.751 0.771 0.763 0.358 0.793 0.799 0.479 0.604 0.168	45.110 43.846 50.146 48.357 48.937 79.794 45.320 44.745 70.756 62.900 91.156	23.504 22.647 25.362 22.429 23.132 37.551 21.168 20.701 34.971 29.319 47.051	0.717 0.752 0.746 0.728 0.734 0.435 0.690 0.741 0.593 0.669 0.023	50.760 48.235 48.390 48.316 48.845 71.182 49.812 49.812 49.391 59.417 56.086 97.894	27.259 24.726 25.518 24.640 25.634 32.648 25.873 25.241 28.658 27.110 52.224
Particulate O	rganic Carbor	1 Concentration						
Power Law (b	iased corrected	d)				Power Law (non-linear least squares)		
			0.474	4.755	1.568	0.584	1.678	1.069
Multiple Linea	r Regression (I	Elastic Net)				Random Forest Regression		
	1 2 3 4 5 6 7 8 9 10 11	$\begin{array}{l} SSC \sim Q \\ SSC \sim Q + H \\ SSC \sim Q + H + P \\ SSC \sim Q + H + P + P_{k,1} \\ SSC \sim Q + H + P + P_{k,1} + P_{k_2} \\ SSC \sim P_{k+1} \\ SSC \sim P_{k+1} \\ SSC \sim P_{k+1} \\ SSC \sim P_{k+1} \\ SSC \sim H \\ SSC \sim H \\ SSC \sim H \\ SSC \sim P_{k+1} \\ SSC \sim P_{k+1} \\ SSC \sim H \\ SSC \sim P_{k+1} \\ SSC \sim H \\ SSC \sim P_{k+1} \\ SSC \sim P_{$	0.362 0.555 0.440 0.592 0.578 0.368 0.464 0.595 0.553 0.515 0.078	1.780 1.506 1.676 1.499 1.526 1.757 1.660 1.485 1.509 1.601 2.006	1.079 0.841 0.974 0.884 0.913 1.197 0.987 0.871 0.839 0.999 1.452	0.467 0.663 0.613 0.642 0.601 0.628 0.503 0.628 0.503 0.479 0.638 -0.263	1.569 1.383 1.356 1.381 1.518 1.598 1.487 1.327 1.522 1.503 2.218	0.874 0.765 0.780 0.784 0.847 0.781 0.866 0.749 0.864 0.749 1.534
Support Vecto	or Regression					Neural Network Regression		
	1 2 3 4 5 6 7 8 9 10 11	$\begin{array}{l} SSC \sim Q \\ SSC \sim Q + H \\ SSC \sim Q + H + P \\ SSC \sim Q + H + P + P_{k1} \\ SSC \sim Q + H + P + P_{k1} + P_{k2} \\ SSC \sim Q + H + P + P_{k1} + P_{k2} \\ SSC \sim P_{k1} + Q \\ SSC \sim P_{k1} + Q + H \\ SSC \sim P_{k1} + Q + H \\ SSC \sim H + P_{k1} \\ SSC \sim P \end{array}$	0.064 0.636 0.663 0.735 0.594 0.498 0.531 0.612 0.541 0.471 -0.022	2.061 1.422 1.343 1.226 1.531 1.648 1.541 1.455 1.579 1.598 2.305	1.144 0.791 0.729 0.670 0.862 0.848 0.797 0.762 0.748 0.792 1.245	0.627 0.579 0.574 0.650 0.285 0.590 0.667 0.656 0.599 0.697 0.188	1.471 1.599 1.411 1.479 1.762 1.699 1.432 1.468 1.468 1.466 1.392 2.250	0.792 0.817 0.760 0.743 0.840 0.887 0.779 0.781 0.713 0.701 1.168





Table B2: Modelled export of suspended sediment and particulate organic carbon using traditional and machine learning approaches, averaged over 47 yr (1974-2020 inclusive).

	Mean annual flux (t)							
		Q < 12.7 m s ⁻¹	Q > 12.7 m s ⁻¹	Summer	Fall	Winter	Spring	
Suspended Sediment								
Power Law (bias-corrected)	25,788.18±3,775.68	1,816.35±48.17 (7.04 %)	23,971.83±3,765.95 (92.96 %)	12,498.98±3,055.6 8 (48.47 %)	3,373.31±615.71 (13.08 %)	3,100.66±500.32 (12.02 %)	6,815.23±2,343.25 (26.43 %)	
Power Law (non-linear least squares)	17,789.77±1,041.51	4,821.24±91.88 (27.10 %)	12,968.53±1,009.22 (72.90 %)	6,502.93±748.95 (36.55 %)	3,061.53±357.74 (17.21 %)	3,023.15±295.25 (16.99 %)	5,202.16±716.82 (29.24 %)	
Multiple Linear Regression (Elastic Net)	25,455.62±1,595.16	5,885.17±146.12 (23.12 %)	19,570.46±1,534.49 (76.88 %)	9,550.15±1,140.49 (37.52 %)	4,241.03±551.88 (16.66 %)	4,174.50±453.24 (16.40 %)	7,489.94±1,092.33 (29.42 %)	
Support Vector Regression	19,673.18±842.56	4,055.42±127.67 (20.61 %)	15,617.76±763.59 (79.39 %)	6,337.29±535.88 (32.21 %)	3,557.99±3,557. 99 (18.09 %)	3,603.13±362.26 (18.31 %)	6,174.77±562.59 (31.39 %)	
Random Forest Regression	25,166.54±1,055.81	5,675.07±193.27 (22.55 %)	19,491.48±931.47 (77.45 %)	8,151.50±627.60 (32.39 %)	4,618.15±522.81 (18.35 %)	4,514.11±431.31 (17.94 %)	7,882.78±752.69 (31.32 %)	
Neural Network Regression	24,854.07±1,741.07	3,841.13±118.92 (15.45 %)	21,012.94±1,689.47 (84.55 %)	9,651.12±1,260.24 (38.83 %)	4,013.30±576.11 (16.15 %)	3,900.18±478.11 (15.69 %)	7,289.47±1,194.11 (29.33 %)	
Particulate Organic Carbon								
Power Law (bias-corrected)	762.68±121.06	45.47±1.24 (5.96 %)	717.21±120.82 (94.04 %)	379.06±98.69 (49.70 %)	96.21±18.21 (12.61 %)	87.57±14.70 (11.48 %)	199.84±74.24 (26.20 %)	
Power Law (non-linear least squares)	426.30±21.39	136.13±2.42 (31.93 %)	290.17±20.46 (68.07 %)	148.80±14.94 (34.90 %)	75.80±8.07 (17.78 %)	75.42±6.63 (17.69 %)	126.29±14.84 (29.62 %)	
Multiple Linear Regression (Elastic Net)	638.48±33.50	200.05±5.52 (31.33 %)	438.44±30.37 (68.67 %)	227.97±22.17 (35.71 %)	109.50±13.67 (17.15 %)	108.35±10.94 (16.97 %)	192.66±22.42 (30.17 %)	
Support Vector Regression	573.48±24.59	156.04±4.89 (27.21 %)	417.44±21.36 (72.79 %)	187.97±14.85 (32.78 %)	104.64±12.58 (18.25 %)	100.16±9.29 (17.46 %)	180.70±17.00 (31.51%)	
Random Forest Regression	539.98±21.56	183.60±5.70 (34.00 %)	356.38±17.61 (66.00 %)	167.94±12.25 (31.10 %)	98.29±10.62 (18.20 %)	98.39±8.60 (18.22 %)	175.36±15.03 (32.48%)	
Neural Network Regression	580.23±32.64	136.21±4.30 (23.48 %)	444.01±30.25 (76.52 %)	209.88±22.26 (36.17 %)	97.41±12.92 (16.79 %)	95.15±10.37 (16.40 %)	177.79±22.19 (30.64%)	

Table D1: Results of significant non-parametric Mann-Whitney and Kruskal-Wallis ranks sum tests as well as Conover-Iman post hoc tests.

Mann-Whitney rank sum test			Kruskal-W	Kruskal-Wallis rank sum test					Conover-Iman post hoc test			
		difference	p-value		df	н	p-value		difference	p-value		
SSC (mg/L)	Baseflow - stormflow	-231.64	<0.001	δ²Η (‰)	3	25.98	<0.001	Fall - Spring Fall - Summer Spring - Summer	4.23 -0.81 -4.89	<0.001 1.000		
DOC-F ¹⁴ C	Baseflow - stormflow	-0.03	<0.001					Fall - Winter Spring - Winter Summer - Winter	-4.03 3.86 -0.54 4.56	0.001 1.000 <0.001		
				8 ¹⁸ O (‰)	3	24.25	<0.001	Fall - Spring Fall - Summer Spring - Summer Fall - Winter Spring - Winter Summer - Winter	4.29 -0.74 -4.88 3.35 -1.08 3.99	<0.001 1.000 <0.001 0.004 0.853 <0.001		
				POC-8 ¹³ C (‰)	3	24.05	<0.001	Fall - Spring Fall - Summer Spring - Summer Fall - Winter Spring - Winter Summer - Winter	2.73 -3.03 -5.33 1.15 -1.67 4.08	0.023 0.010 <0.001 0.761 0.294 <0.001		
				DOC-F ¹⁴ C	3	10.55	0.010	Fall - Spring Fall - Summer Spring - Summer Fall - Winter Spring - Winter Summer - Winter	2.20 3.29 0.91 1.34 -0.99 -2.03	0.092 0.005 1.000 0.556 0.970 0.139		





030



Figure B1: Performance of (a-b) traditional power law, (c) multiple linear regression (MLR), (d) support vector regression (SVR), (e) random forest regression (RFR), (f) neural network regression (NNR) models. The evaluation is based on observed against predicted suspended sediment values (mg L⁻¹). Performance metrics are based on nested cross-validation (R^2 : coefficient of determination, RMSE: root mean squared error, MAE: mean absolute error).







Figure B2: Performance of (a-b) traditional power law, (c) multiple linear regression (MLR), (d) support vector regression (SVR), and (e) random forest regression (RFR), (f) neural network regression (NNR) models. The evaluation is based on observed against predicted particulate organic carbon values (mg L^{-1}). Performance metrics are based on nested cross-validation (R^2 : coefficient of determination, RMSE: root mean squared error, MAE: mean absolute error).



040

