An empirically-derived hydraulic head model controlling water storage Decadal in-situ hydrological observations and outflow over a decade empirical modeling of pressure head in degraded permafrost rock slopes (Zugspitzea high-alpine, D/A)fractured calcareous rock slope

Riccardo Scandroglio<sup>1</sup>, Samuel Weber<sup>2,3</sup>, Till Rehm<sup>4</sup>, and Michael Krautblatter<sup>1</sup>

**Correspondence:** Riccardo Scandroglio (r.scandroglio@tum.de)

1 Abstract

<sup>&</sup>lt;sup>1</sup>Landslide Research Group, TUM School of Engineering and Design, Technical University of Munich, Munich, Germany

<sup>&</sup>lt;sup>2</sup>WSL Institute for Snow and Avalanche Research SLF, Davos Dorf, Switzerland

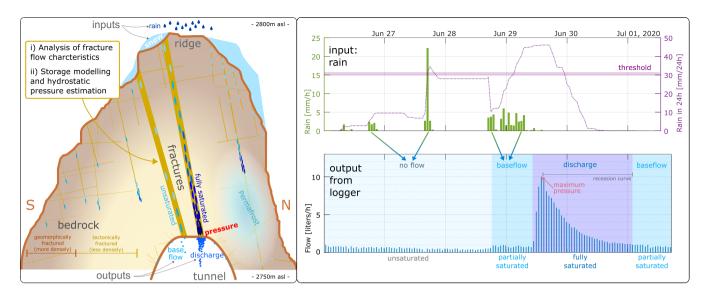
<sup>&</sup>lt;sup>3</sup>Climate Change, Extremes and Natural Hazards in Alpine Regions Research Center, CERC, Davos Dorf, Switzerland

<sup>&</sup>lt;sup>4</sup>Environmental Research Station Schneefernerhaus, Zugspitze, Germany

#### **Abstract**

15

While recent permafrost degradation in Alpine In peri- and paraglacial slopes has been documented in several studies, only restricted information is available on the respective hydrology. Water boosts permafrost degradation by advective heat transport and destabilizes periglacial mountain slopes. Even if multiple recent regions, water plays a critical role in the hydrological cycle and slope stability. However, hydrological models often overlook water infiltration into bedrock due to limited knowledge of groundwater dynamics at high elevations. Although the link between water presence and rock slope failures indicate the presence of water, only a few studies provide evidence of water availability and related hydrostatic pressures at bigger depths. showing a is evident in many cases, proof of hydrostatic pressure buildup at depth is scarce, highlighting another significant research gap. This study combines a unique decennial data set of aims to decipher the hydrological dynamics and empirically derive hydrostatic pressures in deep bedrock. We present unique decennial meteorological data, snowmelt modeling, and discharge measurements from two rock fractures in a tunnel located at  $\approx 55$  m depth under the permafrost-affected N-S facing Zugspitze Ridge (2815-2962 m asl). To decipher the hydrological properties of fractures, we analyze inputs. We developed an empirical hydraulic model and detected flow anomalies by comparing inputs (i.e., snowmelt and rainfall, and outputs.) and outputs (i.e., discharge from fractures, baseflow, and no-flow events, detecting flow anomalies. For summer precipitation events, we developed i) a uniform recession curve, ii) an empirical water storage model, and iii) an approximate hydraulic water pressure model according to Darcy's falling-head law). Results show continuous flow during snowmelt and discontinuous events during summer months. Hydraulic conductivities are in the order of  $10^{-4}$  m/s, with variations according to the saturation. Extreme events with are likely to fully saturate the fractures and increase their interconnectivity, producing discharges up to 800 l/d and 58 l/h are likely to fully saturate the observed fractures with corresponding hydraulic heads of up to 40 from one single fracture. Hydrostatic pressures calculated implementing Darcy's falling-head law are  $27 \pm 10$  m and to increase fracture interconnectivity. The average daily discharge during snowmelt, 10 l/h, can lead to hydraulic heads up to 27 6 m during average snowmelt and  $40 \pm 6$  m. Water dynamics suggest hydraulic conductivities in the range of  $10^{-4}$  m/s, with variations according to the fracture's saturation, E.g., no flow and baseflow events indicate unsaturated and partially saturated conditions. Here, we show an empirical fluid flow approximation model of hydrostatic pressure regimes in high-alpine deep-bedrock fractures. Pressures from water accumulation in bedrock reach levels that m for extreme events. These pressure levels can weaken or even destabilize rock slopes. This process can easily outpace thermal conductive warming of active layers in the foreseeable future, provide positive feedback on water infiltration, and is crucial for the stability of the rapidly warming alpine permafrost environments in rapidly warming alpine environments. With ongoing climate changes, water relevance is expected to increase, with impacts that yet have to be fully assessed. This study advances the understanding of alpine hydrology and geomorphology by providing new insights into deep groundwater processes and their implications for slope stability.



. Graphical abstract. Left: relevant geomorphological and hydraulic components. Right: input and output measurements.

#### 1 Introduction

35

High mountain regions are recognized as "global water towers" (Viviroli et al., 2007), which massively sustain seasonal water availability for ~ 1.9 significantly contributing to water resources for approximately 1.9 billion people (Immerzeel et al., 2020), but knowledge of groundwater dynamics in high-alpine rock slopes is searce. Most studies focus on sub-alpine watersheds with soil and vegetation covers and infer groundwater dynamics only through spring discharge (Hayashi, 2020) due to the lack of hydrological measurements at high elevations. Logistically challenging terrain, harsh meteorological conditions, and substantial pattern variability (Walvoord and Kurylyk, 2016; Arenson et al., 2022) are limiting factors in high alpine environments. So far, only a few studies have directly monitored groundwater in alpine hillslopes with deep wells in competent and fractured bedrock (Manning and Caine, 2007; Gabrielli et al., 2012). The models applied often assume that bedrock in alpine catchments behave like "Teflon basins" with flow only at the surface (Clow et al., 2003) unless additional sediment-rich surface layers are available.

Developments suggest that deep fractures are a crucial pathway for groundwater flow along the hillslope and that this can be highly dynamic (Banks et al., 2009). Steady baseflow during periods of little recharge indicates the relevance of aquifers (Hayashi, 2020), contributing 5—. During periods of limited recharge, alpine groundwater becomes particularly important, providing up to 50% to adjacent lowlands aquifers (Markovich et al., 2019), seasonally redistributing water, and stabilizing eatchment outflow (Cochand et al., 2019). Geological conditions mainly control groundwater in mountain bedrock and can be divided into two components. On the one hand, the main flow component is shallow and is topographically driven because of the higher weathering in the first meters (Clarke and Burbank, 2011; Welch and Allen, 2014). On the other hand, the deep bedrock flow is characterized by a complex and heterogeneous permeability reduction with depth (Manning and Caine, 2007)

, and of adjacent lowland aquifers (Markovich et al., 2019; Hayashi, 2020). Groundwater flow within mountain bedrock can be categorized into two distinct components. The primary flow occurs at shallow depths and is predominantly driven by topography (Clarke and Burbank, 2011; Welch and Allen, 2014). The secondary flow is deeper and exhibits greater complexity: it is controlled, among others, by fracture density, geometry, and connectivity, which can be highly complex and heterogeneous (Banks et al as well as a decrease in permeability with depth (Manning and Caine, 2007).

Permafrost is known to influence groundwater flow paths and storage (Woo, 2012) and allows the accumulation of perched water table above the frozen material (Krautblatter et al., 2013). Still, it remains unclear to what extent thawing mountain permafrost contributes to the water cycle, e.g., through groundwater storage (Walvoord and Kurylyk, 2016). Mountain permafrost hydrology studies mainly focus on unconsolidated sediments (Hayashi, 2020), since the majority of alpine springs discharge comes from talus While measurements with lysimeters focus on shallow flow (Courtin and Bliss, 1971; Rist and Phillips, 2005) , there is a considerable lack of knowledge about deep groundwater dynamics in alpine slopes and their connection with the cryosphere (van Tiel et al., 2024). Direct measurements of deep groundwater in fractured alpine bedrock using wells are limited to a few studies (Manning and Caine, 2007; Gabrielli et al., 2012) due to the logistical challenges of rugged terrain, harsh weather conditions, and significant spatial variability (Walvoord and Kurylyk, 2016; Arenson et al., 2022). Piezometric measurements in fractures or boreholes are mostly of difficult interpretation (Draebing et al., 2017; Phillips et al., 2023; Bast et al., 2024) , while combined geophysical methods can identify liquid water presence (Hauck et al., 2011; Watlet et al., 2018; Pavoni et al., 2023) , but cannot measure pressure. So far, studies on high-alpine hydrology have predominantly focused on talus slopes, moraines, moraine, or rock glacier (Noetzli and Phillips, 2019; Jones et al., 2018; Arenson et al., 2022). Only glaciers (Hayashi, 2020; Noetzli and Phillips, 2019; Jones et al., 2018; Arenson et al., 2022). and only recently, Ben-Asher et al. (2023) combined field measurements and numerical modeling to simulate hydrological fluxes on modeled surface hydrological fluxes in steep bedrock permafrost. However, much work is still to be done to understand bedrock hydrology in periglacial areas. From the slope stabilitypoint of view, permafrost bedrock failure results from many factors, e.g., joint sets geometry, presence of a fault zone, or glacier retreat (Haeberli et al., 1997; Hasler et al., 2012; Phillips et al., 2017) . The presence of water in the scarp after rock slope failures, as well as the occurrence of high water availability prior to the failure, has been observed recently in various events (Stoffel and Huggel, 2012; Fischer et al., 2010; Walter et al., 2020; Kristensen et al., 2 , suggesting that high hydrostatic pressure contributes to failure. According to Montgomery et al. (2002), near-surface fractured 75 bedrock can influence landslide triggering positively or negatively, depending on pressure head build-ups or storm runoff

In these environments, water presence also influences slope stability, as shown in numerous failure events (Fischer et al., 2010; Stoffel and where hydrostatic pressure has been hypothesized as one of the destabilizing factors. In the theoretical model by Krautblatter et al. (2013), hydrostatic pressures can lead to increased pressure can destabilize slopes by increasing lateral shear stresson the rock mass, reduce, decreasing frictional strength, and lower effective normal stressleading to a reduction in shear resistance of rock-rock contacts. Active-layer deepening due to climate change increases the infiltration depth of water, generating higher hydrostatic pressures (Haeberli and Gruber, 2009). Recent slope-stability models with and without permafrost focused on the reducing normal stress. Mechanical simulations at slope scale confirm the critical role of water in fractures, proving that it plays a crucial role in slope within fractures for stability (Scandroglio et al., 2021; Magnin and Josnin, 2021). From the thermal

accommodation. In the rock-/ice-mechanical-

point of view, water percolating in bedrock fractures can quickly thaw fractures bedrock by advection and, therefore, destabilize bigger larger rock masses than heat conduction (Haeberli et al., 1997; Gruber and Haeberli, 2007), but. Still, so far, direct field evidence of water availability and consequent sudden thermal disturbance at depth is only available for one site (Gemsstock (CH), Phillips et al., 2016)(Phillips et al., 2016).

Different approaches to detect and quantify water presence range from multi-method geophysics in permafrost (Hauck et al., 2011; Wagn and in unfrozen karst (Watlet et al., 2018), over piezometric measurements in frozen cracks (Draebing et al., 2017) and rock glacier (?Bast et al., 2024), to surface lysimeters (Courtin and Bliss, 1971; Rist and Phillips, 2005). So far, all methods have shown substantial limitations in bedrock, and water pressure at deep depthshas not been directly measured yet.

Despite the recent increasing importance and interest in this research field, we still lack measures and understanding of how deep bedrock groundwater (> 10 m) couples to superficial water availability on a short-term (days) and long-term (seasonal) scale in alpine environments. Inferring water accumulations and their destabilizing potential remains a core challenge that requires further research. In this study, we combine a decade of underground discharge measurements, weather data, and snow simulations to derive Despite the critical role of fracture water in alpine environments, there is a lack of reliable data and modeling efforts, resulting in an inadequate understanding of groundwater flow dynamics and little knowledge of hydrostatic pressures at depth. To address these two gaps, we present a decade-long dataset of high-alpine fracture discharge measurements at significant depths. First, we investigate the spatial and temporal patterns of water flow in deep bedrock fractures in response to snowmelt and rainfall events. Then, we develop an empirical model of water dynamics in the first decameters under the surface ( $\approx 55\,\mathrm{m}$ ) of fracture water dynamics to constrain catchment characteristics and identify periods of water accumulation. Finally, we provide innovative estimates of hydrostatic pressures in a high-alpinelow-porosity bedrock. We address the following questions:

i)How can we estimate fluid flow's spatial and temporal behavior in near permafrost bedrock fractures? ii)How can we use fracture depletion to constrain fluid flow curves? iii)How can we use outflow curves to constrain water storage? iv)Can we generate a Darcy falling head model to mimic and generalize fluid flow in bedrock fractures?

, fractured calcareous rock slope.

90

100

105

110

#### 2 Study sitedescription and characterization

Measurements took place on Mount Zugspitze (2962 m asl, Fig. 1a), located in the Northern Calcareous Alps at the German-Austrian borderand visited by thousands of tourists daily. The study area site is located southwest of the summit, on in a no longer used pedestrian tunnel under the east-west oriented ridge between Zugspitze and Zugspitzeck. A disused tunnel for pedestrians runs from southwest to northeast under this ridge (Fig. 1b), and it is Loggers are placed at approximately 2750 m asl and are accessible all year round thanks to its direct connection to from the research station Scheefernerhaus (UFS).

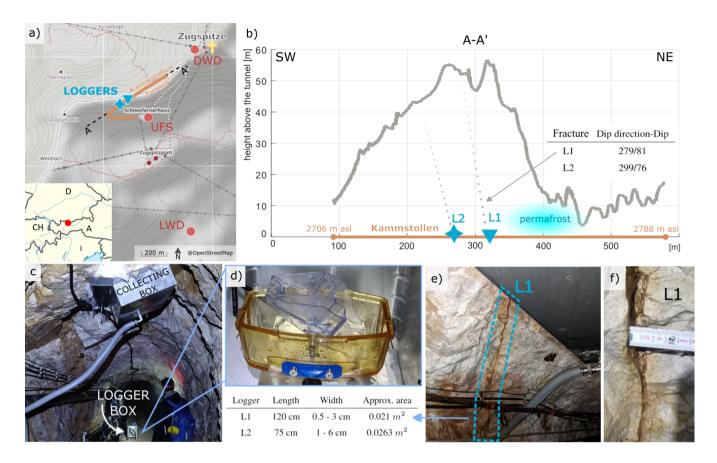


Figure 1. Overview of the study site. a) Inset map: location of the Zugspitze. Main map: summit (in yelloweross, 2962 m asl) with the location of the Kammstollen tunnel in orange. Blue signs represent the two flow loggers, and red circles the weather stations: DWD = German Meteorological Service, UFS = Environmental Research Station Schneefernerhaus, and LWD = Bavarian Avalance Service. ©OpenStreetMap contributors 2023. Distributed under the Open Data Commons Open Database License (ODbL) v1.0. b) Section A-A' of the ridge along the tunnel showing shows the distance between the tunnel and the surfaceoverlaying rock masses. The tunnel ascends from 2706 to 2788 m asl; the loggers are at approx. 2750m asl. The location of the loggers table and the permafrost lens (Krautblatter et al., 2010) is shown. The table presents dotted lines present the orientation dip direction and dip of the two fractures instrumented with loggers. c) The Measuring setup: collecting box is installed on the ceiling, connecting pipe, and is connected to the logger with a pipebox. d) Tipping gaugewith resolution 0.1 L. e) and table with measures f) Fracture L1 and detail of both fracturesit with scale. The approximate Table with fractures' dimensions: the fracture area covered by the sample collecting box is length by width.e) Fracture L1 and d) detail of L1 with scale.

# 2.1 Climate and cryosphere

Long-term meteorological records by the German Meteorological Service (DWD) have existed since 1900 A long-term meteorological record on the summit (Fig. S1 in the supporting materials). The mean temperature in the last decade (2013-2022)was -3.3°C, which is 1.5°C warmer than the reference period 1961–1990, with 2022 being the warmest year ever recorded, with -2.7°C

on averagehas existed since 1901 by the German Meteorological Service (DWD). The climate is influenced by the prominent elevation at the northern edge of the Alps and by multiple E-W oriented ridges, leading to mean annual precipitation of more than  $2500 \,\mathrm{mm}$ , with no recent changes compared to the reference period. 80% of precipitation is snowfall from autumn to late spring, most parts of the catchment are snow-free at the end of July, and heavy thunderstorms, as well as long-lasting, intense rainfall events, may occur during the summer season (Wetzel et al., 2022). The mean temperature in the last decade  $(2013-2022) \,\mathrm{was} -3.3 \,\mathrm{^{\circ}C}$ , which is  $1.5 \,\mathrm{^{\circ}C}$  warmer than the reference period 1961-1990.

Mayer et al. (2021) documented the dramatic recession of the three glaciers at the Zugspitze: with no mass increases since the 80s, the glaciers are destined to disappear in the next decade. Permafrost occurrence and degradation are well documented and closely monitored at the summit (Gallemann et al., 2017) and in the tunnel (Krautblatter et al., 2010). The permanently frozen bedrock of the tunnel is located Glacier and permafrost degradation in the area has been extensively documented (Mayer et al., 2021; Gallemann et al., 2017). A permanently frozen lens has been detected a few decameters from our study site. Geophysical the water loggers (Krautblatter et al., 2010). Monthly geophysical measurements are conducted monthly to monitor the evolution of the permafrost lens and detect possible interaction with percolating water. An ice-controlled slope instability has been observed on the ridge, 400m from our site, and it was monitored recently by Mamot et al. (2021) with permafrost geophysical mapping and deformation monitoring, to monitor its dynamics and identify potential interaction with infiltrating water.

# 2.2 Hydrology

135

140

150

The basin south of the summit, also known as Research Catchment Zugspitze (RCZ, 11.4 km²), with its high-density sensors network is one of the best instrumented high alpine high-alpine catchments for monitoring hydrological processes thanks to a high-density sensors network over four elevation zones and hydrograph recording at the downstream Partnach spring (Wetzel, 2004; Weber et al., 2021). Hydrochemical investigations by Rappl et al. (2010) and Weishaupt (2021) using natural environmental tracers and electric conductivity provided the catchment borders and their hydrogeological characteristics, which prove the catchment to be defining it as a perfect natural lysimeter. They evidenced a karst water reservoir in the phreatic zone beneath the Zugspitze Plateau that can hold around half of the volume of summer precipitation. During the winter season, from the end of October to April, no karst system recharge occurs, and Partnach-Spring the outflow spring falls dry (Morche et al., 2008). Because of this dynamic, climate-change-induced variations in the snow cover will strongly affect water availability in the RCZ and in the areas downstream (Weber et al., 2016). Furthermore, Voigt et al. (2021) used relative gravity measurements to detect water storage variations in the RCZ at the catchment scale with promising results. Still, the only known measurements of water discharge in shallow unsaturated bedrock were conducted for measuring persistent organic pollutants in shallow percolated water (Levy et al., 2017), using the same loggers as in this study.

#### 2.3 Geology and fractures

The whole summit area is composed of Triassic Wetterstein limestone, with a thickness of about 600–800 m (Hornung and Haas, 2017). Ulrich and King (1993) report brecciated zones up to 1 m thick that dip steeply  $(60^{\circ}-90^{\circ})$  in the directions of

NW–ENE and can be intercalated with ice. A relevant fault zone can be found from above the UFS up to the summit, and karst dissolution is frequent, especially on the Plateau. The Krautblatter et al. (2010) mapped the fractures in the tunnel were mapped in 2007 Krautblatter et al. (2010) and are here newly analyzed in Figure S2 of the supplementary material. Fractures with a dip of 80-90° are the majority (S1, n=41) but without a predominant direction (SD = 90°). Few other discontinuities are mapped, mainly with orientation 58/106(S2, N=7)and 49/311(S3, N=11). This The two fractures where the loggers are installed also belong to this group, as confirmed by punctual measurements conducted by Georg Stockinger in 2023 (personal communication). The indoor mapping agrees only partially with the one from Mamot et al. (2021), which conducted scan lines and field mapping from the surface 400m NW from our site (Fig. S2e). In addition, punctual measurements on the two main fractures, where the water gauges are installed, were conducted by Georg Stockinger in 2023 and are presented in Figure 1bThis difference is mainly due to the influence of the fault zone.

### 3 Methods Data and datamethods

160

175

In this study, we present a modeling approach to estimate the hydraulic head (hereafter also called hydrostatic pressure), revealed from rain and snowmelt time series (system input) and measurements of water discharge from fractures in the tunnel (system This section introduces our model's main components, which are water entering the system (fracture inputs) and water leaving the system (fracture output). Using this information, we decipher the fluid flow in deep bedrock fractures (system process)Definitions and methods used for analyzing water dynamics are then presented in the last part of this section.

# 170 3.1 System Fracture input: water from rain and snowmelt

Inputs to the model can be rainfall data from the weather stations or snowmelt data from the software Snowpack. Meteorological Snowpack. The meteorological measurements are conducted at three locations, shown in Figure 1 and presented the three locations listed hereafter and in Table 1.

- i) DWD The German Meteorological Service records air temperatures every 10 minutes and the daily precipitation on the summit of the Zugspitze. Due to the exposed location of this weather station, results are often influenced by winds and northern weather.
  - ii) UFS The Environmental Research Station (2650 m a.s.l.) measures precipitation and air temperatures every 10 minutes together with the DWD. In the middle of the south slope, this location is protected from the northern winds but exposed to western atmospheric perturbations.
- 180 iii) LWD The Bavarian Avalanche Service runs an automated snow and meteo station on the Zugspitzplatt (Plateau, 2420 m asl). This station is in a central and flat position, protected from strong south, west, and north winds. Values are recorded with 10 minutes resolution.

For the snowmelt modeling, we use the one-dimensional open-source software Snowpack (SP)that models, which simulates the evolution of the snow cover based on weather data . It simulates (Lehning et al., 1999). It reproduces mass and

Station	Parameter	Resolution	Time analysed	Usage
DWD	$TA, \stackrel{\mathbf{P}}{\underset{\sim}{\sim}}$	yearly, monthly, year, month	1901-2023	long-term PSUM (hourly, daily) trends
UFS	TA, P	10 Min.	2000-2023	rainfall
PSUM modelling height LDW	HS. TA	10 Min.	2000-2023	snowTA-Snowpack modelling

**Table 1.** Metereological input data used in this study. Stations: DWD = German Meteorological Service, located on the summit and often influenced by winds; UFS = Environmental Research Station, located at 2650 m asl in the middle of the south slope, is protected from the northern winds but exposed to western atmospheric perturbations; LWD = Bavarian Avalanche Service snow and weather station, located on the plateau at 2420 m asl in a protected and flat position. The exact locations are visible in Figure 1a. Parameters: TA = Air air temperature, PSUM-P = sum of precipitation, HS = snow height.

energy exchanges taking place between the atmosphere, snow, and soil and reproduces the evolution of snow microstructure (Lehning et al., 1999). Simulations were . Simulations are conducted with data from LWD for every hydrological year separately, with 15-minute time steps. The provided inputs inputs provided are incoming and outgoing shortwave radiation, snow depth height (HS), relative humidity, air temperature (TA), sum of precipitation (PSUM), temperature of the snow surface snow surface temperature, and wind speed/direction. The measured snow depth is a proxy for precipitation inputs to force the mass balance. Data gaps are interpolated with the integrated pre-processing library MeteoIO. Ground temperature is set constant at  $0^{\circ}$ C, and albedo is estimated from based on incoming and reflected shortwave radiation. Boundary conditions for snow melting are adapted each year to fit the melting phase best, but due to model limitations, discrepancies between modeled and measured snow heights are still possible. The most important output is the amount of meltwater  $[kg/m^2]$  that leaves the snow cover in liquid form and reaches the ground.

## 195 3.2 System Fracture output: water discharge from fractures

205

Two water collecting systems, L1 and L2 (Fig. 1c-d and Fig. S3 in the supplementary material), were installed in 2010 in the central part of the pedestrian tunnel to collect water that leaks from two independent fractures (POPALP Report, 2011; Levy et al., 2017). Each system is composed of one comprises a collecting box installed on the ceiling of the tunnel(Fig. S3 in the supplementary material) tunnel's ceiling, a connection pipe, one and a logger box. Inside this, a tipping bucket with a reed sensor for measuring the discharge, and one logger for data recording. The resolution of 0.11, a reed sensor, and a logger record the discharge hourly. The collecting boxes are located in the unfrozen area, as confirmed by electrical resistivity tomography resistivities (Krautblatter et al., 2010), and their vertical distance to the surface obtained with a DTM is  $\approx 55 \,\mathrm{m}$ . The fractures corresponding to each logger and their measures are shown in Figure S3 (see supporting material). Discharge 1e-f and S3. The UFS has provided discharge data since 2013are available thanks to the UFS staff; new loggers were installed in 2020.

### 3.3 System process: fluid flow in deep-bedrock fractures Analysis of fracture water dynamics

The system inputs and outputs are united for analyzing and modeling fluid flow. We first perform a systematic analysis to characterize it, then model water flow and accumulation in fractures based on (i) recession curves analysis, (ii) detection of flow anomalies, and (iii) assessing of fracture saturation. Finally, we suggest an approach to quantify the hydraulic head.

### 3.3.1 Systematic analysis

210 Discharge measurements from snowmelt or rainfall events, meteorological data, and SPSnowpack results were united are combined in Matlab at hourly and daily resolution. Snowmelt While snowmelt is analyzed seasonally and daily, while rainfall is only on an event basis. Snowmelt produces uninterrupted flow as water flows uninterrupted for many weeksup to months, making it hard to define single events. Meanwhile, rainfall is naturally event-based, and often-, rainfall is categorized into events as dry periods clearly mark the limits. Mixed events are excluded, while whereas rain-on-snow cases in spring are included. An output flow event starts with a sudden increase of the discharge, independent of the starting value, and 215 ends when the flow returns to a value smaller than a threshold (typically 1 l/h). Baseflow is a constant or very slowly decreasing here defined as an almost constant discharge of small magnitude (typically < 1 l/h) that happens after (or before) a flow event and can last up to some weeks, even without further input. In our model presented in Section 5.3, baseflow is closely connected to fracture saturation. By convention, multiple flow events are classified as one if precipitation interruptions are shorter than 24 220 hours and if the resulting hydrograph at the gauges does not reach baseflow status between the two rain events. Input-output anomalies in the flow are detected and analyzed. One example is no-flow events when rain generates no relevant flow in the fracture. After a manual pre-selection, timing and quantities are analyzed. The rain events are selected manually and present a high heterogeneity of quantities and duration. Analyses are conducted automatically with a dedicated Matlab function where parameters are that uses variable parameters provided by the user. They are variable from caseto case due to the heterogeneity of the events for each case. All selected events are listed in Figure S7 of the supplementary material, together with the input-225 output graphics for each event in Figure S8 and S9. Some analyses are limited to fracture. The analysis includes 23 rain events from L1 since L2 recorded less events.

#### 3.3.1 Model fluid flow and water accumulation in fractures

We only model rain events because they can be clearly separated from each other thanks to the dry phases in between. Since input and output quantities are all directly measured, the resulting empirical model is certain. On the contrary, snowmelt is the output of SP modeling; therefore, quantities there must be considered a likelihood and not a precise measure. Additionally, during snowmelt, water flows uninterrupted for days to weeks, making it harder to detect single events.

#### Recession-curve analysis

230

235

The recession-curve analysis represents a basic hydrogeological research tool, used for over a century but still up-to-date, with modifications and expansions. Recession curves can be fitted to drainage data to fully reproduce the runoff with empirical coefficients. This provides information on the flow characteristics and on the attributes of the aquifer, e.g., estimation of

karstification degree or groundwater sensitivity to pollution (Malík and Vojtková, 2012; Kirchner, 2009). Boussinesq (1877) was the first to describe aquifer drainage and spring discharge through a porous medium using a diffusion equation. Using simplifying assumptions, he obtained the approximate analytical solution described by exponential Equation 1, where  $Q_0$  is the initial discharge,  $Q_t$  is the discharge at time t, and  $\alpha$  is the recession coefficient, an intrinsic aquifer parameter.

$$Q_t = Q_0 e^{-\alpha t}$$

240

245

250

255

260

265

Maillet (1905) also obtained similar results for reservoir emptying through a porous plug: knowing initial discharge values and recession coefficients, the entire runoff curve can be described. Hydrogeologists widely use Equation 1 due to its simplicity and linearisation in logarithmical plots, but other recession equations are also available to better fit various shapes of hydrograms. For example, linear equations are used for fast-flow components typical from karstic channels (Malík and Vojtková, 2012). Complex aquifers with mixed flow regimes (e.g., with karst and multiple conduits) require a combination of different equations to reproduce their groundwater circulation. One or more laminar and turbulent sub-regimes may exist in one aquifer, and the total discharge can be described by the superposition of all the flow equations.

We took all the events and extracted the recession part of the discharge curve. If one event shows multiple peaks, the event has been divided into sub-events, and one curve has been assigned to each peak. All the curves are first plotted starting at time t = 0 (Fig. ??a in the supplementary material). Then, recession curves are analyzed from the biggest to the smallest and each curve is shifted in time so that its starting value fits the same value of the bigger event (Fig. ??b). We computed the mean of these shifted curves and tested different fittings to the mean curve, according to the procedure explained in Malík (2015). A function that is the sum of several exponential segments (Eq. 2) is used to represent the complete recession hydrograph of a karst spring.

$$Q_t = \sum_{i=1}^{n} Q_{0i} e^{-a_i t}$$

#### Flow anomalies detection

During data collection, unusual situations were observed where the presence or absence of water in the tunnel could not be directly explained. To understand these cases, we analyze all possible situations in Figure ??: after separating between snow-covered and snow-free periods, we considered snowmelt (SM), precipitation (PSUM) and air temperature (TA), listing all possible combinations. Finally, we highlighted unexpected flow behaviors in orange and looked for reasonable explanations: all these situations could be explained by water accumulation in the bedrock layer above the tunnel. Flow anomalies detection.

a) Simplified model representing all the components of hydrological significance. Storage includes fractures, karst voids, and pores, i.e., all locations where water can accumulate. In the tunnel, we find the collecting box (C) and the logger (L). b) Detection of unexpected water behavior in the tunnel; the upper case represents snow-covered periods, and the lower one represents snow-free periods. Abbreviations: SM = snowmelt, PSUM = sum of precipitation, TA = air temperature. Green boxes show standard situations, while the red boxes highlight situations requiring water accumulation within the rock mass. Possible accumulation places within the bedrock, hereafter abbreviated as *storage*, are (i) karst voids, (ii) pores, and (iii) fractures. i) The

presence of karst voids in the region of the Zugspitze Plateau is well documented (Wetzel, 2004), but their presence in the area above the tunnel has not been proven yet. Karst would also strongly influence the flow behavior, and it is, therefore, unlikely. ii) An average rock matrix permeability of 5 μD (Krautblatter, 2009) limits the amount of water that can flow through the rock. iii) Fractures, due to their filling with fine material, are preferred flow paths and can temporarily store water in accumulation periods. Therefore, we focus our model on this last component.

# Fracture saturation and storage model

Here, we present an empirical model that explains flow anomalies with fracture saturation and storage variations (Fig. ??).

The model is composed of four stages: from unsaturated (S1) to partially saturated (S2a/b) to fully saturated (S3). Repeated alternation of steps 2b and 3 is possible when multiple precipitation events happen close enough in time. During dry periods, S1 resumes again until the next precipitation. Flow events can only happen if fractures are fully saturated, while baseflow is the outcome of partially saturated fractures. The main outputs of the model are i) a qualitative forecast of the saturation level in the fracture and ii) a quantitative estimate of the amount of water present in the fractures at each time step.

**Fracture saturation model.** Upper part: four phases of our model. Lower part: exemplary summer rainfall event. Precipitation (blue bars) from UFS, discharge (green line) from Logger 1, and the corresponding fill level of the storage in the fracture from Eq. 4 (red line). Point A) represents the maximum storage in the fracture system, and point B) represents the storage at maximum discharge. Table with steps of the fracture saturation model as a consequence of input and output.

For ii), we first calculated the cumulative quantities at the end of the event: the discharge (Q) in litersand the precipitation (PSUM) in mmor  $I/m^2$ . While the first is exactly known for every event, the exact amount of precipitation infiltrating the fracture is unknown. Therefore, for each event, we use Equation 3 to calculate the ratio  $R_{end}$  that expresses the surface required to obtain the measured flow in the tunnel.

$$\underline{R_{end} = \sum_{t=0}^{end} Q \, / \sum_{t=0}^{end} PSUM}$$

Since we excluded superficial runoff and evaporation, this value is only a minimum estimation.  $R_{end}$  is then used in Equation 4 to compute the storage fill level  $FL_t$  at each time step t, that is, the difference between cumulative input and output.

$$FL_t = \sum_{0}^{t} PSUM * R_{end} - \sum_{0}^{t} Q$$

#### 3.3.1 Conversion in water column

Finally, we estimate the resulting hydraulic head using Darcy's Law (Eq. 5), the basic equation that describes fluid flow through saturated porous media.

$$q = -K\frac{\Delta h}{\Delta l}$$

Here, yq is the specific discharge  $[LT^{-1}]$ , K is the hydraulic conductivity  $[LT^{-1}]$ , h is the head [L], and l is the travel distance [L] (Zha et al., 2019). According to Bernoulli's equation (Eq. 6), the total hydraulic head  $h_t$  can be composed of elevation head  $h_z$ , pressure head  $h_p$ , and velocity head  $h_v$ . In this case,  $h_v$  can be neglected due to extremely low velocities in porous media, and  $h_p$  should be overall the same. Therefore, the elevation head  $h_z$  is the dominant component and it drives water flow.

$$h_t = h_z + h_v + h_p \approx h_z$$

300

305

310

To constrain the hydraulic head, it is reasonable to compare our case to a falling-head test with a Darcy cylinder as in Figure ??a, where the only constant head is at the discharge point (for example in Allan Freeze and Cherry, 1979). Application if Darcy's falling head test. a) Darcy cylinder with falling head in the pipe above and constant head at the outflow. b) Illustrative S-N transect of the fracture for water flow with falling head. Bedrock deeper than 8 meters is expected to be less densely fractured (Clarke and Burbank, 2011). Suppose the fracture is simplified with a cylinder having a constant diameter over the whole length (a = A, in Fig. ??b). In that case, L is the "effective path" that offers hydraulic resistance, and therefore, a hydraulic head can build up above it. In Figure ??a, the discharge according to Darcy is  $Q = qA = -K \cdot A \cdot H/L$  and it must be equal to the change in hydraulic head, i.e., the discharge rate above the effective path  $Q = a \cdot dh/dt$  recorded in the last 10 years, while it was possible to detect a relevant flow in only half of the cases for L2 (Tab. S7, supplementary material).

$$a\frac{dh}{dt} = -KA\frac{h}{L}$$

Since here a = A we can simplify, rearrange the terms and integrate the equation, obtaining:

$$\int\limits_{h_1}^{h_2} \frac{dh}{h} = -\frac{K}{L} \int\limits_{t_1}^{t_2} dt$$

$$ln(h_2) - ln(h_1) = -\frac{K}{L}(t_2 - t_1)$$

If we solve for the maximum event length, we suppose  $t_1 = 0$  and  $t_2 = t_{max}$ , obtained from the recession curve. Considering the baseflow at the end  $h_2 = 0.1$  m, we can compute  $h_1$ , the hydraulic head at the beginning.

$$h_1 = 0.1 \cdot e^{\left(K \cdot t_2/L\right)}$$

### 4 Results and data interpretation

Water flow is categorized into

# 4 Results and interpretation

The first two sections of this chapter evaluate groundwater dynamics of spring snowmelt and summer rainfall, evaluated separately this chapter's first two sections separately. The third section then focuses on no-flow events, baseflow, and extreme events, while the following sections compute the recession curve and analyze the storage. After defining some premises, the last section estimates the hydraulic heads, linking them with discharge values, which are crucial for the empirical model.

# 4.1 Snowmelt induced discharge

Daily (a) and hourly (b) snowmelt in spring 2023 is shown in Figure ??, while all 2. All seasons are available at daily resolution in the supplementary Figure S4of the supplementary material. Figure ?? 3 presents snowmelt statistics from 2013 to 2023.

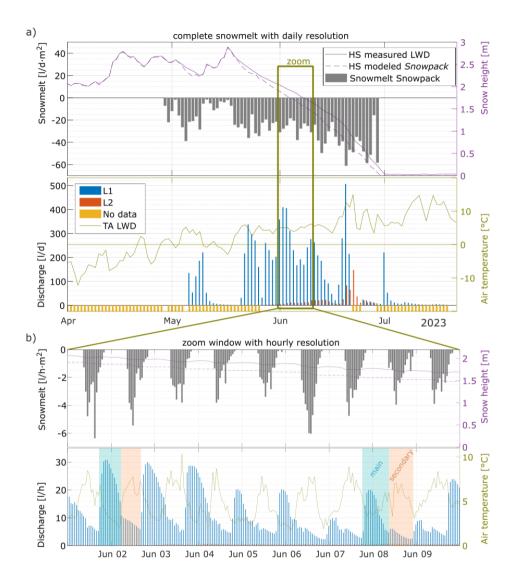


Figure 2. Example of modeled snowmelt rates and measured water discharge for summer 2023, Example of modeled snowmelt rates and measured water discharge for summer 2023, daily rates in subfigure a) and hourly rates in subfigure b). The top panel of each subfigure shows the water input into the rock, i.e., melting water and snow height. The bottom panels show the water output, i.e., the discharge measured in gauges L1 and L2 (almost absent < 1 l/h in this easethe zoom window), and air temperature. In the bottom panel of b), two different phases of water flow are highlighted: the main flow in blue and the secondary flow in red.

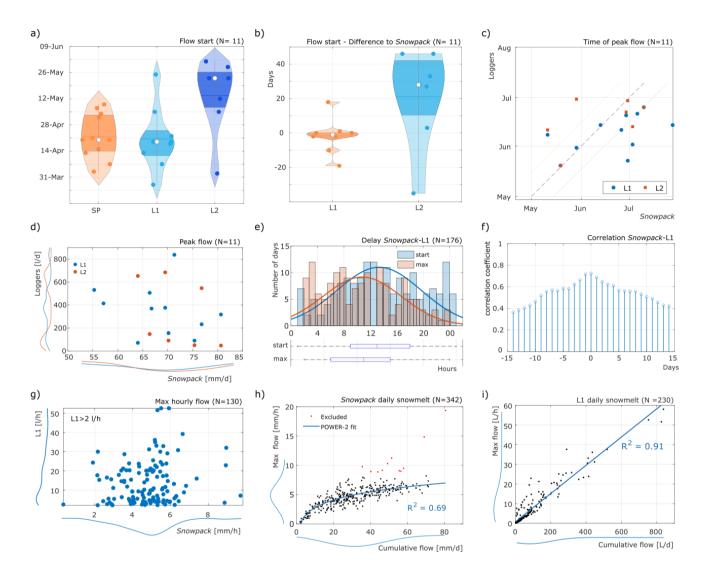


Figure 3. Snowmelt statistics. a) Violin plot of the flow start. The white points represent the median, and the darker areas represent the 25th and 75th percentiles. b) Delay between the Snowpack Snowpack (SP) model and fluid flow loggers. c) Time of maximum flow for SP compared to L1 and L2. The line represents the same day. d) Maximum daily discharge for each year for L1, L2 and Snowpack, with probability distribution. e) Delay between SP and gauge L1 for flow begin and maximum discharge. f) Daily correlation between SP and L1, with the peak at 0 and -1 days. g) Maximum hourly flow for SP and L1 for each day. h) Maximum daily melting rate and correlated daily cumulative melting for SP. Excluded exceptions mostly represent rain-on-snow events. i) Maximum daily fracture flow and correlated daily cumulative flow for L1.

**Seasonal and daily analysis -** Snow melting Snowmelt generally starts at the end of April and lasts until the end of June (Fig. ??2a). Small events are also possible in summer or autumn, mostly with negligible snowquantities. Daily values flow rates present good temporal agreement between measurements and model for Snowpack (SP) modeling concerning the starting

and the pauses interruptions, but discharge stops earlier than in the model (e.g., in 2019, 2021, and 2022). The daily modeled melting rates vary strongly across the season: values increase with time, reaching the maximum at the end. Snowmelt starts in the second half of April for the SP The start time of snowmelt is very similar for the model and for L1, while L2 starts in is delayed to the second half of May (Fig. ??a). SP and L1 differ only by one day, while L2 has 3a), with a median delay of 28 days from SP (Fig. ??3b). Yearly peak snowmelt is concentrated Peak snowmelt for SP is mainly at the end of Junefor SP, while both loggers reach their peak about two weeks before or earlier (Fig. ??3c). Still, 5 cases show a very good fitting in time. The yearly peaks Yearly peak values reach 80 mm/d for SP and 840 l/d for L1, but no correlation between variables is evident (Fig. ??3d). The stronger variations variability of L1 and the fact that, on average, its flow is five times higher than SP can be explained by increased fracture interconnectivity for periods of high flow, so more surface contributes to these events.

Hourly analysis - Figure ??b highlights some disagreement between 2b shows measured and modeled values at hourly resolution. Some disagreements are evident due to water travel time, storage effects, and possibly model limitations. According to the SPmodelSP, snowmelt occurs only during daily hours, from 4:00 to 21o'clock:00 local time, with a maximum at 13 o'clock and no flow during the night (Fig. S5a in the supplementary material). As expected, melting hours per day increased towards the end of June (Fig. S5b in the supplementary material): this explains the maximum daily melting rates of that period. On the contrary, according to L1 and L2, water flows continuously in the tunnel with daily cycles that vary in intensity and timing over the season, which can be divided into two phases (Fig. 22b). The main flow during night hours, marked in blue, and the secondary flow during daily hours, marked in red. The superimposition of hydrographs can explain these two flows: water is coming from at least two paths that have different lengths, fracture apertures, and/or filling compositions. In fracture L1, the The end of an event corresponds to the start of the next. New: new events mostly start around 12 and have a maximum at 22 o'clock (Fig. S5c). The delay between the beginning of SP melting and water flow increase in the tunnel is 13 h, while the delay of the between peaks is only 11 h (Fig. ??3e), but both show substantial variability. Here, we cannot exclude that the actual delay is a multiple of the calculated one. To verify this, we plotted the correlation between SP and L1 for daily values over 5 years (Fig.  $\frac{??f}{?}$ ; for each year separately 3f and Fig. S6 in the supplementary material). A maximum correlation of > 0.7is reached at 0 and 1 day, therefore, Therefore, the delay of peaks can be 11 h or 11+24 h. Variable delays This variability can be explained by increased fracture interconnectivity or changes in the fracture saturationlevel or the interconnectivity infill saturation: with high saturation, the hydraulic conductivity increases, and water flows faster. The daily maximum flow rate for SP is 4.7 mm/h on average, with a maximum at 9 mm/l, while the values of h, while for L1 are is 10 l/h on average, with a maximum at 58 l/h (Fig. ??3g and i). No correlation is evident here. The daily cumulative flow of SP does not fit the hourly maximum flow rates linearly (Fig. 23h), while the daily total flow of L1 is linearly correlated to maximum flow rates (Fig. ??3i). This is because the quantity of water that can be released by 1 m<sup>2</sup> of snow during one day is physically limited, but with higher discharges, more fractures are connected with higher discharges.

#### 4.2 Rainfall induced discharge

345

350

355

360

This analysis is based on distinct rain events manually selected with high heterogeneity of quantities and duration. L1 recorded 23 rain events in the last 10 years, while it was possible to detect a relevant flow in only half of the cases for L2 (Tab. S7,

supplementary material). All possible flow trajectories after resulting from a rain event are explained in Figure ??.4. The output discharge from L1 and L2 can differ even with the same input, e.g., in Event 2 (yellow area) a flow is recorded only in L2, indicating different saturation pre-saturation levels in the two fractures. For Event 3, on the contrary, discharge is recorded in both loggers, therefore, both fractures are fully saturated.

365

370

375

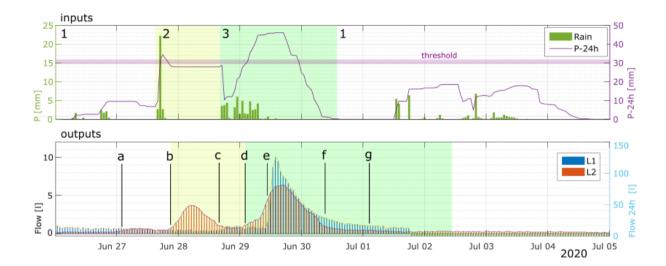


Figure 4. Example of a period with multiple summer precipitation events (i.e., inputs) and the corresponding flow in the fractures (i.e., outputs). Abbreviations: PSUM P - sum of hourly precipitation, PSUM-24h P-24h - sum of precipitation in the last 24 hours. The upper-Upper panelshows the inputs: 1) low-intensity short-duration event with PSUM-24 P-24 smaller than the threshold (≈ 30 mm/24 h), 2) high intensity-short duration event around the threshold, and 3) low-intensity long-duration events with PSUM-24 P-24 above the thresholdand 4) again like 1). The lower Lower panelshows the hourly outputs: L1 in blue and L2 in red. a) L1: no changes, L2: increase of baseflow, b) L2: begin of event, c) L1: increase of baseflowflow, d) L2: begin of the second event, e) L1: begin beginning of the first event, f) L2: end of the event, only baseflow, g) L1: inflection point, end of the event, only baseflow.

Results from the 23 events show that the duration of flow events in the tunnel is mostly longer than precipitation duration (Fig. ??a), but 5a). Still, the correlation between duration in L1 and L2 is very high (Fig. S10 of the supplementary material). Even short rain events cause flow events (< 10h) cause discharges that last longer than two days. There is a sudden increase in flow duration for longer more prolonged precipitation (> 80 mm), which could signify higher also points to an increase in fracture interconnectivity. Flow in the gauge L1 can last up to 7 times longer than the precipitation (Fig. ??5f). The delay from precipitation start to flow in the tunnel is 31 h for both loggers (Fig. ??5b), but this time can vary ± 10 h according to the amount of precipitation occurring in the previous 3 days (Fig. ??5c). This value is a good proxy for the pre-saturation level of the fracture: the more rain before the event, the higher the fracture's saturation, so the faster the water flows. Maximum hourly flow rates in L1 (3.9 l/h) are smaller than precipitation (8 mm/h) due to slow percolation smoothing the peaks (Fig. ??5d), but if we consider the 24h-sums, the discharge in the tunnel becomes proportionally bigger larger than the precipitation (Fig. ??5f)

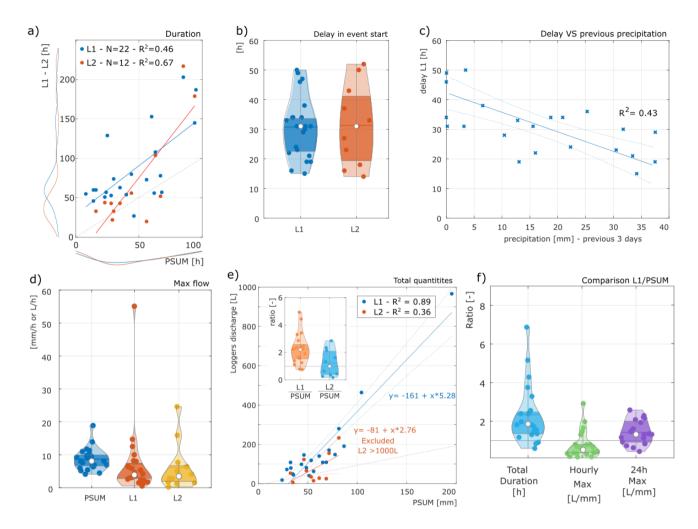


Figure 5. Analysis of rain events. a) Correlation for event duration: PSUM-P compared to flow in L1 and L2. b) Violin plot of the delay between precipitation and flow in L1 and L2. c) Relation between precipitation in the three days before the event and delay for L1. d) Violin plot with maximum hourly precipitation and flow rate. e) The main graph shows the correlation analysis for the total quantities, L1 or L2 vs. PSUMP. The small graph shows the rations L1/PSUM-P and L2/PSUM-P for total quantities. f) Ratios L1/PSUM-P for different quantities: event total duration, maximum flow rate, and 24h-cumulative maximum flow.

and Fig. S10). Total quantities confirm this: more than one liter of water reaches the tunnel for each mm of rain falling on the surface: 2.3 l/mm of rain in L1, 1 l/mm for L2 (Fig.  $\ref{eq:25}$ e). These values  $\ref{eq:25}$ can be converted in m<sup>2</sup>  $\ref{eq:25}$ can be co

#### 4.3 No-flow , baseflow and extreme events

385

390

395

400

410

No-flow - Some rain events Rain events that do not produce a relevant water flow in the tunnel , and they are hereafter defined as no-flow events. We selected 49 summer rain events to investigate this anomaly (Tab. S2, supplementary material). No-flow events last up to 36 h and have a have maximum precipitation of  $\approx$  34 mm. The median of the 6h-sum of precipitation and last up to 36 h. Still, the 6h-precipitation is very similar to the total precipitation, meaning that the precipitation is mostly concentrated in a few hours. Total precipitation and peak intensity are poorly correlated with duration (Fig ??6b), but total precipitation is related to the peak intensity (Fig ??6c), confirming the predominance of short high-intensity rainfalls. No events show long duration and low intensity, indicating that rainfall generating no-flow events is mainly short or has high-intensity, likely thunderstorms. This is further supported by the absence of no-flow events that exhibit long durations and low intensities.

**Baseflow** - The importance of baseflow is well highlighted in Figure ??, where baseflow precedes and forecasts the arrival of an event in the tunnel. Baseflow is also fundamental in our model (Fig. ??): during S2b, baseflow is explained by a partially saturated fracture. During dry summer periods, baseflow reduces to zero if two precipitation events are far enough from each other, as in Sweetenham et al. (2017). In this case, we suppose the fracture to be unsaturated.

**Extreme events -** The logger recorded two special cases with extreme discharges. The first is the snowmelt on 10th-12th June 2019, which happened after a record snow depth of more than 6 m in Mayand in connection, coincidently with a sudden increase in air temperature up to +10°C. This caused extreme tunnel discharges up to 800-750 l/d for three consecutive days. The second extreme event is the rainfall on 16th-18th July 2021, with peak intensities of 160 mm/ 24 h and 20 mm/h, preceded by multiple smaller rain events in the five days before, which pre-saturated the fractures. This event was forecasted by the public warning service and generated floods in the valley. In the tunnel, discharges reached values above 800 l/d and hourly values up to 55 l/h, the maximum in the last 10 years. In this case, the delay between peak precipitation and peak flow is was only 3 hours.

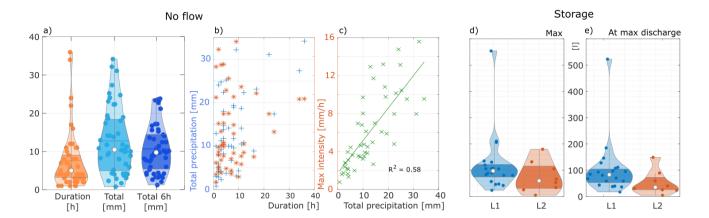
#### 4.4 Recession curve fitting

405 Comparing all the L1 discharge curves normalized from 0 to 1 in time and quantity shows that they follow

## 5 Modelling flow and storage in deep-bedrock fractures

In the following section, we compute the recession curve and analyze flow anomalies. Afterward, we explain the anomalies by introducing the empirical model for fracture saturation and quantifying stored quantities. Finally, discharge values are used to estimate hydraulic heads. The following analyses focus solely on rain events because they are distinctly separated by dry periods, and all input/output quantities are directly measured and, therefore, are highly reliable.

#### **5.1** Recession-curve fitting



**Figure 6.** Left: analysis of no-flow events AAnalysis of no-flow events and storage a) Statistics and spread for 49 events. b) Correlation between duration and total precipitation (blue crosses) or maximum intensity. c) Correlation between total precipitation and maximum intensity. Right: analysis of storage. d) Maximum storage and e) storage at maximum discharge.

Recession curves can fully reproduce discharge behavior using empirical coefficients. These aquifer intrinsic parameters provide information on the flow characteristics and on the attributes of the aquifer, e.g., estimation of karstification degree (Malík and Vojtková, 2012; Kirchner, 2009). Boussinesq (1877) was the first to describe aquifer drainage and spring discharge through a porous medium using a diffusion equation. Using simplifying assumptions, he obtained the approximate analytical solution described by Equation 1, similar to Maillet (1905) for reservoir emptying through a porous plug.

$$Q_t = Q_0 e^{-\alpha t} \tag{1}$$

Here  $Q_t$  is the discharge at time t,  $Q_0$  is the initial discharge, and  $\alpha$  is the recession coefficient. Other equations are available to better fit various shapes of hydrograms, e.g. linear equations. Complex aquifers with mixed laminar-turbulent flow regimes, like karst with multiple conduits, require more equations to fully reproduce their groundwater circulation (e.g., Eq. 2).

420

425

$$Q_t = \sum_{i=1}^n Q_{0i} e^{-a_i t} + \sum_{i=1}^m Q_{0i} (1 - \beta_i t)$$
(2)

We first analyzed all events in L1. Normalized curves show a similar pattern (Fig. S11 and S12 of the supplementary material) . On average, the discharge peaks after the first with a discharge peak after only 10% of the time. Then, within only 30Later, at 40% of the time, it discharge decreases back to 30% of the peak, and the remaining 60% of the time is required to return to baseflow.

We apply the recession analysis to the events from L1, as shown in Figure ?? in the supporting material. The best fitting is obtained with two exponential components, i.e., For the recession-curve fitting, we extracted the discharge part from all events and divided multiple peaks into sub-events, with one curve for each peak. All the curves are first plotted at time t=0 (Fig. 7a) and then, starting from the biggest values, each curve is shifted in time so that its starting value fits the same value

of bigger events (Fig. 7b). Finally, we computed the mean of the shifted curves and tested different fitting solutions for it (Fig. 7c), according to Malík (2015). The best results for our spring are obtained using Equation 2 with n = 2, that and

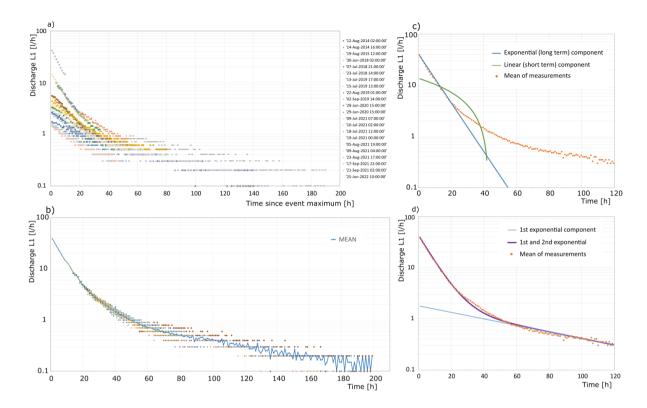


Figure 7. Modelling of all discharge events with a uniform recession curve. a) All events are plotted with  $t_0 = 0$  and semi-logarithmic axes. b) Discharge curves are shifted in time so that  $Q_0$  fits a similar value from a higher curve. In blue is the mean curve. c) Recession curves fitting with single equation linear flow component (green line) and single equation exponential flow (blue line). d) Recession curve fitting with 2-components exponential flow.

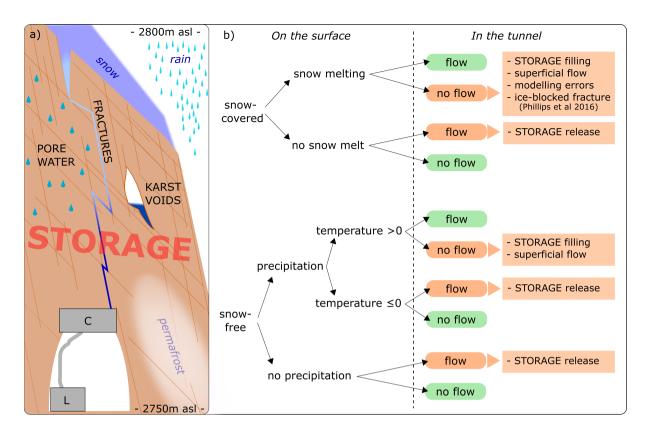
m = 0 (Fig. 7d), which reduces the error to an acceptable level (Table S1 in the supplementary material). According to the classification suggested by Malík and Vojtková (2012), "the combination of two or more sub-regimes with merely laminar flow characterized by different discharge coefficients and (exponential equations) characterized by higher values of  $\alpha_x$ .  $\alpha$ " describes a flow happening in an "aquifer with-irregularly developed fissure network, with the majority of open macrofissures, and with the possible presence of karst conduits of limited extent".

#### 5.2 Flow anomalies detection

435

440

Combining input on the surface and output in the tunnel, we observe situations where the presence or absence of water in the tunnel can not be directly explained. These unforeseen flow behaviors are analyzed in Figure 8, and all are explainable by water accumulation in the bedrock above the tunnel ("storage"). Possible storage includes (i) karst voids, (ii) pores, and



**Figure 8.** Parameters Flow anomalies detection. a) Simplified model of all hydrological components. Storage includes fractures, karst voids, and pores, i.e., all locations where water can accumulate. In the mean recession curve fitting tunnel, we find the collecting box (C) and the logger (L). b) Analysis of water presence: green boxes show expected situations, while orange boxes highlight unexpected situations with possible explanations. All these cases can be explained by water accumulation in the bedrock.

#### Fitting Flow component $\alpha_x$ (or $\beta_x$ ) 1

(iii) fractures. i) Results of the recession curve confirm that the presence is possible but only to a limited extent. Extensive karst voids are present under the Plateau (Wetzel, 2004), but their presence in our study area has not been proven yet. Therefore, we exclude this possibility. ii) An average rock matrix permeability of 5  $\mu D$  (Krautblatter, 2009) strongly limits the storage possibilities in the matrix. iii) Fractures are preferred flow paths and can temporarily store water due to their filling with fine material. Therefore, our model will focus on this component.

#### 5.3 Fracture saturation and storage model

445

450

Figure 9 presents the empirical model that explains flow anomalies. The model is composed of four stages and allows a qualitative forecast of the saturation level in the fracture: from unsaturated (S1) through partially saturated (S2a and S2b) to fully saturated (S3). S2a and S2b differentiate for the outflow: in S2a, we have no flow, while in S2b, we have baseflow. Discharge in the tunnel happens when fractures are fully saturated (S3). S2a/ $\frac{hQ0_a}{a}$  b and S3 can alternate repeatedly when

precipitation events happen in quick succession. During dry periods, when S1 resumes, and baseflow reduces to zero, we suppose the fracture to be unsaturated, as in Sweetenham et al. (2017).

This model allowed us to estimate the amount of water stored in the fractures at each time step. For this, we calculated the cumulative discharge Q [I/h] Error<sup>1</sup> Linear fast-flow and precipitation P [I/turbulent  $\beta_1 = 0.023$   $Q0_1 = 13.6$  4103 | Exponential 155 | 1 slow-flow / laminar  $\alpha_1 = 0.109$   $Q0_1 = 41.9$  86 | Exponential 2 slow-flow / laminar  $\alpha_1 = 0.015$  /  $\alpha_2 = 0.123$   $Q0_1 = 1.8$  /  $Q0_1 = 41.4$  17 | 1

### 5.4 Empirical storage modeling

460

470

As next step, we use  $m^2$ ] at the end of the event. While the first is exactly known, the exact amount of rain infiltrating the fracture is unknown. Therefore, for each event, we calculated the ratio  $R_{end}$  (Eq. 3), which is the surface required to obtain the discharge measured in the tunnel. This value represented only a minimum estimation since we excluded superficial runoff and evaporation.

$$R_{end} = \sum_{t=0}^{end} Q_t / \sum_{t=0}^{end} P_t \tag{3}$$

 $R_{end}$  was then used in Equation 4 to compute the fill level FL of the storage determine the difference between cumulative input and output at each time step -t, which corresponds to the storage level  $S_t$ .

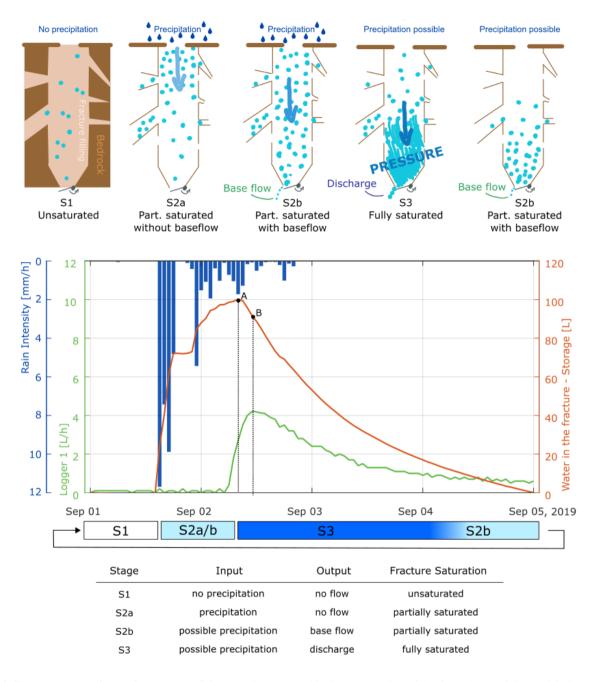
465 
$$S_t = \sum_{t=0}^{t} P_t * R_{end} - \sum_{t=0}^{t} Q_t$$
 (4)

The supplementary material presents the resulting storage curves for all events in Figure S13 and S14. The maximum water distributed average storage in the fracture system is on average 97 l for L1 and 59 l for L2 (Fig. ??6d). Since

Considering that the timing of the maximum storage does not fit with the maximum discharge, we suppose that water reaches the maximum pressure when the maximum discharge is different from the time of maximum discharge in the tunnel is recorded (see points A and B in Fig. ??)and 9), we expect the maximum pressure in the fracture when the maximum discharge is recorded. Here, we also suppose that all the water in the fracture is concentrated in at one point, which is the bottleneck of the fracture. At maximum discharge, the average storage Storage reduces by 15-20%, but the highest value still from A to B, and the maximum storage at B reaches 520 l in July 2021 (Fig. ??6e).

### 475 5.4 Estimating hydrostatic pressures from dischargemeasurements

<sup>&</sup>lt;sup>1</sup>as absolute difference



**Figure 9. Fracture saturation and storage model.** Upper image: graphical representation of the four phases of the model. Central graph: exemplary summer rainfall event. Precipitation (blue bars), fracture discharge (green line), and the corresponding storage level from Eq. 4 (red line). Point A represents the maximum storage in the fracture system, and point B represents the storage at maximum discharge. Lower table: stages of the fracture saturation model as a consequence of input and output.

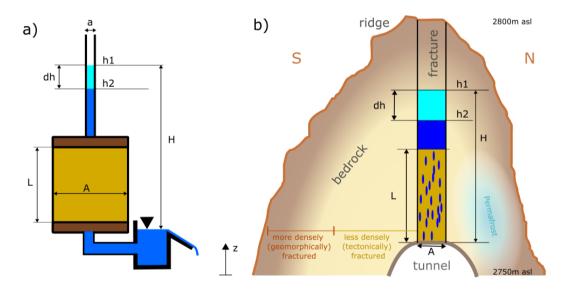
Finally, we estimated the resulting hydraulic head using Darcy's Law (Eq. 5), the basic equation that describes fluid flow through saturated porous media.

$$q = -K\frac{\Delta h}{\Delta l} \tag{5}$$

Here, q is the specific discharge  $[LT^{-1}]$ , K is the hydraulic conductivity  $[LT^{-1}]$ , h is the head [L], and l is the travel distance [L] (Zha et al., 2019). According to Bernoulli's equation (Eq. 6), the total hydraulic head  $h_t$  can be composed of elevation head  $h_z$ , pressure head  $h_p$ , and velocity head  $h_v$ . In this case,  $h_v$  can be neglected due to extremely low velocities in porous media, and  $h_p$  should be constant within the system. Therefore, the elevation head  $h_z$  is the dominant component, and it drives water flow (Allan Freeze and Cherry, 1979).

$$h_t = h_z + h_v + h_p \approx h_z \tag{6}$$

To constrain the hydraulic head, we compared our case to a falling-head test with a Darcy cylinder as in Figure 10a and made the following four assumptions. i) The only constant head is at the discharge point. ii) The diameter is constant over the whole length (a = A, in Fig. 10b). iii) L is the "effective path" that offers hydraulic resistance; therefore, a hydraulic head can build up above it, iv) Lateral flow is neglected.



**Figure 10. Application of Darcy's Law for a falling head.** a) Darcy cylinder with falling head in the pipe above it and constant head at the outflow. b) Illustrative S-N transect of the fracture for water flow with falling head. Bedrock deeper than 8 meters is expected to be less densely fractured (Clarke and Burbank, 2011).

With the information obtained so far and applying Equation 9, we constrain a realistic value range for  $h_1$ , the maximum 490 hydraulic headat the beginning of a discharge eventIn Figure 10a, the discharge according to Darcy is  $Q = q \cdot A = -K \cdot H/L \cdot A$  and it must be equal to the change in hydraulic head, i.e., the discharge rate above the effective path  $Q = a \cdot dh/dt$ . We apply the following premises.

$$a\frac{dh}{dt} = -KA\frac{h}{L} \tag{7}$$

Since here a = A, we can simplify, rearrange the terms, and integrate the equation, obtaining:

495 
$$\int_{h_1}^{h_2} \frac{dh}{h} = -\frac{K}{L} \int_{t_1}^{t_2} dt \qquad ln(h_2) - ln(h_1) = -\frac{K}{L} (t_2 - t_1)$$

$$(8)$$

Solving for the maximum event length, we suppose  $t_1 = 0$ , and  $t_2 = t_{max}$  is obtained from the recession curve. Considering baseflow at the end ( $h_2 = 0.1 \,\mathrm{m}$ ), we can compute the hydraulic head at the beginning ( $h_1$ ).

$$h_1 = 0.1 \cdot e^{(K \cdot t_2/L)} \tag{9}$$

The following boundary conditions apply in this case.

510

- 500 i) An absolute boundary condition comes from the The height of the erestline surface above the tunnel:  $h_1 < 55 m$ .
  - ii) General bedrock porosity, including matrix and fractures, is estimated at  $2.5 \pm 1.5$  % (Krautblatter, 2009), which gives an average storage capacity of  $25 \pm 15$  l for each  $m^3$  of rock. Considering the extreme event of July 2021, 5201 must be stored above the tunnel, resulting in  $13 < h_1 < 52$  m. These numbers might be smaller, as we know that since, on the surface, porosity increases, and more fractures are presentand fractures increase.
- 505 iii) The length of the effective path, L, must be realistic: 5 < L < 20 m. We fix one value for the whole process.
  - iv)  $\frac{t_2}{t_2}$  is the The theoretical length of the maximum event and  $\frac{t_2}{t_2}$  can be obtained from the recession curve analysis:  $t_2 = 200 \text{ h.}$
  - v) A plausible range of k-hydraulic conductivities is obtained from literature and from the recorded events. According to Allan Freeze and Cherry (1979), hydraulic conductivity—K in karst limestone can vary between  $10^{-6}$  and  $10^{-2}$  m/s. The travel time of water is known, i.e., start delay, and the flowing path is supposed to be the same as the distance surface-tunnel. The resulting, while estimated field velocities are  $\approx 5 \cdot 10^{-4}$  m/s.

Given With the given assumptions, these boundary conditions and applying, and Equation 9to compute  $h_1$ , a plausible range for, we constrained a realistic range for the hydraulic head at the beginning of an event  $h_1$ . First, plausible couples of K and L values are were obtained (Fig. ??11a and Fig. S3 of the supplementary material). These couples are were then validated for different event lengths (75 <  $t_2$  < 200 h) in Figure ??11b. A realistic value of K appears appeared to be between  $5 \cdot 10^{-5}$  and  $1.5 \cdot 10^{-4}$  m/s, we choose chose the median  $1 \cdot 10^{-4}$  m/s, which requires an effective length L between 11.5 and 12.5 m to

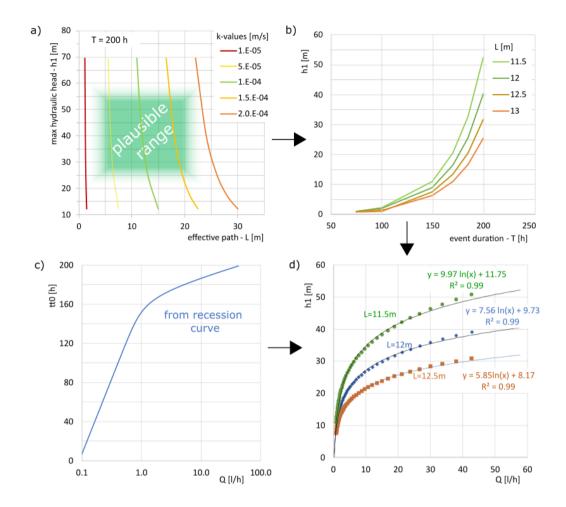


Figure 11. Approximative hydraulic head model. a) Hydraulic head resulting from different effective path (L) and hydraulic conductivity (K). The area in green highlights physically realistic results. b) Validation of the results by changing the event duration. c) *Time to 0 flow* (tt0) for each discharge Q. d) Hydraulic head H1 computed using discharge from fracture L1. For this case, we used  $k = 10^{-4} \ m/s$ .

produce a max hydraulic pressure h1 between 32 and 52 m. The validation of these results for shorter event duration confirms their feasibility. confirmed their feasibility (Fig. 7b). Each discharge requires a specific time to return to baseflow after an event (Fig. ??b, supplementary material). This is here. This was defined as time to 0 flow (=tt0), and can be was obtained for any given discharge Q with Equation 2 and the parameters in Table S1. The resulting values of tt0 (Fig. ??e) are 7c). The resulting values of tt0 were used in Equation 9 to compute h1 (Fig. ??11d). This way, we connect connected discharges from the logger and hydraulic heads in the fracture. The fitting logarithmic curve for Q > 1 l/h produces  $R^2 = 0.99$ .

520

Considering the average maximum discharge from the rain events,  $Q \approx 4$  l/h, a hydraulic head of 20 m  $\pm$  4 m can be is obtained. For snowmelt, we have an average recorded daily  $Q \approx 10$  l/hthat can generate, which generates a hydraulic head of

525 27 m ± 6 m. Extreme snowmelt in June 2019 could generate a hydraulic head of 40 m ± 10 m in the fracture. Similar values could have been reached after the, and similar values are also possible for intense rainfall in July 2021.

## 6 Discussion

530

535

540

545

550

555

This article provides valuable insight into the dynamic of water flow insights into water flow dynamics, reservoir effects, saturation levels, and pressure effects in recently degraded permafrost and saturation levels in bedrock fractures. It introduces an empirical method for quantifying hydrostatic pressures generated by snowmelt and rain infiltration. Differently from previous studies (Scandroglio et al., 2021; Magnin and Josnin, 2021), here, for the first time, we quantify pressures not based on modeling results but we dont't quantify pressures from model results but based on a decade of underground discharge measures in high alpine fractureshigh-alpine underground discharge measurements. The robustness of these flow data is the strength of this studyunique flow data indicates the occurrence of periods with high water pressure, allowing the following extensive discussion.

## 6.1 Snowmelt- and rain-driven water flow dynamics in deep fractures as system input

Snowmelt modeled average Average and maximum daily infiltration rates from snowmelt are ≈ 30 mm/d and 80 mm/d, respectively, and are similar to those measured by Rist and Phillips (2005). The software SP-Snowpack can reproduce the timing of extreme melting events but shows disagreement at the end of the melting phase, due to the poor performance of the model model's poor performance or to the different locations of the snow station. More snow is available on the flat Plateau compared to the 40/50°-steep slopes, and regular avalanche detachments for safety reasons further reduce snow availability. A basal ice layer at the cold interface rock-snow is supposed to strongly limit snowmelt penetration in rock walls (Phillips et al., 2016; Ben-Asher et al., 2023). Still, Kneisel et al. (2014) documented thermal disturbance in the underground as soon as snowmelt started, while Kristensen et al. (2021) and Roth and Blikra (2009) recorded large increases in rockslide displacements late in the snowmelting season. Our direct flow measurements show that snowmelt infiltrates every year and that the timing of discharge discharge timing for fracture L1 fits with SP-Snowpack modeling from the Plateau, which lies southward. This confirms that suggests that i) the catchment of L1 is on the southern slope and that a ii) no basal ice layer is not blocking meltwater here. In contrast, a temporary basal ice layer or frozen fractures cannot be excluded for L2, given its starting delay in some years. A seasonal basal ice layer could also be present on the steep northern slopes, which are impossible to reach in winter and where flow where snowmelt happens later in the season, as observed during fieldwork.

Summer precipitations reach the tunnel only for events above  $\approx 30$  mm/24 h. Poulain et al. (2018) confirm that a saturation threshold is necessary to allow vadose connectivity, although values depend from site to site (Sweetenham et al., 2017). High-intensity short-duration events like thunderstorms often happen at the beginning of precipitation and mainly generate, generating no-flow or only baseflow. Most water does not infiltrate, and instead, it of this water drains away as surface flow due to the total absence of a substrate that acts as a buffer. For the same reason, evaporation effects can be neglected at this location. These effects could influence the estimated storage, so we might overestimate it. The amount of precipitation in the period

before the an event influences the travel time of water in bedrock because the fracture so that pre-saturated fractures have a higher hydraulic conductivity *K*. Zhou et al. (2006) found that infiltration Infiltration rates and fracture conductivity correlate with fracture patterns at the inlet (Zhou et al., 2006). Therefore, detecting the precise inlet location could further improve the understanding of the hydrologic system. Unfortunately, the ridge is very exposed, and access is possible only at high risk.

#### **6.2** Outflow as the summative system output

560

565

570

585

The response hydrographs of the two fractures are different: L2 presents a symmetrical hydrograph with similar rising and falling limbs (Fig. ??4d to f), while for L1 the rising limb is very steepbut. Still, the falling limb has a slow decay (Fig. ??4e to g). The differences could be due to different fracture filling, catchment shapes, isolated karst areas, or higher connectivity. Fracture density also strongly influence influences the flow at depth: scarcely fractured networks have a slightly faster response to precipitation than a denser network (Sweetenham et al., 2017). The total annual outflow for L1 reached the decennial maximum value in 2019 with 8500 l/year, which means a theoretical average of 23 l/d. That year, one But in that year, One single snowmelt period even recorded a total of 2300 l in 3days, days, which is 27 % of the total, showing that extreme events dominate water dynamicshere. Measured maximum flow rates are 4 to 10 times higher than those computed by SP Snowpack and measured by Rist and Phillips (2005) because of a bigger measurement depthand of an increased connectivity of the fracture system our data are recorded at 50 m depth, where fracture connectivity increases during high discharges (Sweetenham et al., 2017). Maximum The highest discharges from snowmelt were reached recorded at the end of the melting season, corresponding to June at this location. For the same period, Weber et al. (2017), Etzelmüller et al. (2022), and Leinauer et al. (2024) demonstrated a clear correlation between snowmelt and increased displacement of unstable slopes. One single rainfall event produced extreme values similar to snowmelt but for a shorter time. A contribution to flow from the neighbor permafrost bodies thaw is theoretically possible. The amounts discharges would be small and visible in dry periods at the end of summer, but it is hard to find signs of it in our data.

### 6.3 Pressurised water inside fractures as an agent driving slope instabilities

The computed hydrostatic pressures strongly depend on the selected parameters, although according to our premises, only limited pairs of hydraulic conductivity K and effective length L are reasonable. The presented values are computed for  $K=10^{-4}$  m/s and L=11.5-12-12.5 m, which are the most likely parameters, but the same results could be obtained using, for example,  $K=5\cdot 10^{-5}$  m/s and L=5.75-6-6.25 m. Still, other K-L couples could lead to different hydraulic pressures. Much higher hydraulic conductivities are measured on the Zugspitzplatt by Rappl et al. (2010) with tracers, but this is due to the well-developed karst system present at that location , which is not present on the slope and not at our site.

In extreme cases, hydrostatic pressures up to  $40 \text{ m} \pm 10 \text{ m}$  can be reached, similarly to the models of Magnin and Josnin (2021). Average values reach 20 m during summer rainfall events and 27 m during spring snowmelt, pressures that can be mechanically critical (Scandroglio et al., 2021). These levels are achieved many times in the summer season and every day during snowmelt, generating repeated loading-unloading cycles that have been rarely considered but can be a crucial destabilizing factor for slope instabilities. Leinauer et al. (2024) and Helmstetter and Garambois (2010) reported that every drop of water

can accelerate or trigger instabilities. This can be true <u>during snowmelt or</u> only for superficial movements since our no-flow measurements and the models of Sweetenham et al. (2017) show that minor rain events do not reach depths of 25-50 m.

Out The proposed model provides qualitative estimates of fracture's saturation level at depth, which is erucial to understand permafrost evolution and also crucial for rock wall destabilization patterns (Magnin and Josnin, 2021). When fractures are fully saturated, we must also consider other destabilizing effects like destabilization acts due to the reduction of cohesion and friction of the fracture's filling material and the reduction of shear strength by counteracting the normal stresses (Scandroglio et al., 2021). During snowmelt, water flows uninterrupted for many weeks, and fractures remain longer saturated, while in saturated for longer periods. In summer, rain events alternate dry periods, so fracture saturation is highly variable, and destabilizing effects last shortly.

## 6.4 Error sources and uncertainties

590

595

605

610

620

600 Error sources and uncertainties are possible in the components of our fracture flow model: the inputs, the outputs, and the estimation of storage and pressure.

Water flow characteristics along the rock fractures

InputInputs: Due to the elevation of the study site<del>and our elassification</del>, the number of snow events (N>100) is much bigger larger than the number of rain events (N=23), which influences statistics quality. Snowpack modeling is computed only in 1D at a different elevation than the ridge: 2D or 3D modeling could improve the fitting of the melting phase. Rainfall events are analyzed only hourly, while a 10-minute resolution would provide better insight into high-intensity, short-duration events. Due to their nature, extreme events are rare and statistically less represented.

**Outputs:** Both loggers suffered repeated failures due to lightning strikes, battery problems, and maintenance. Therefore, data gaps could be mistaken for no-flow events or for ice sealing the fracture. L1 and L2 don't always behave similarly ; e.g., discharge hydrographs differ, but this analysis focuses (e.g., different discharge hydrographs), but the analyses here focus mostly on L1 because only a few events were recorded by L2. In fact, the latter is more prone to failure and shows variations in peak discharge with time that are not clearly explainable.

SystemEstimation of storage and pressures: Calibration Model calibration took place using one extreme rain event that could present higher K and higher fracture interconnectivity than normal events. To include this variability, all events are incorporated in the recession curvethat connects discharge to event length. We have also chosen. We chose robust estimates and performed sensitivity propagation to check the robustness of the results. Bedrock deeper than 8 meters is expected to be less densely fractured (Clarke and Burbank, 2011), and a conductivity reduction up to 65% is expected close to a tunnel due to stress increase and joint closure (Fernandez and Moon, 2010). These effects can strongly influence pressure locally (Montgomery et al., 2002). K varies according to fracture pre-saturation in time according to saturation levels (Fig. ??e); 5c), while porosity and hydraulic conductivity are, very likely, not uniform in space, but for. For simplicity, we don't include these variations in our model. Due to the increase of interconnectivity for high discharges, we cannot exclude that in extreme events other than assumed, water spreads laterally, which and produces hydraulic heads smaller than computed suggested here.

### 6.5 Future changes in the systemOutlook

Snowmelt is expected to begin up to one month earlier by the end of this century (Vorkauf et al., 2021). If so, melting rates will be slower due to the reduced solar radiation early in the year (Musselman et al., 2017), leading to a partial reduction of snowmelt infiltration rates and so consequently lower hydrostatic pressures in fractures during snow-covered periods. Snow-free periods might increase and so and the number of summer rainfall events might increase. Heavy precipitation will generally become more frequent and more intense with global warming (IPCC, 2023). Accordingly, we expect i) more events with liquid precipitation and ii) more frequent and more intense extreme flow events in fractures, resulting in higher hydrostatic pressures. Due to climate change, the recently enhanced permafrost degradation increases active layer thickness, creates in snow-free periods.

The impact of permafrost bodies on flow paths (Woo, 2012) and water accumulation (Krautblatter et al., 2013) has largely been overlooked. However, climate change is expected to increase active-layer depth, leading to the formation of new horizontal and vertical flowing paths, and affects massively fracture permeability. In fact, flow pathways. Consequently, the infiltration depth of water will increase, generating higher hydrostatic pressures (Haeberli and Gruber, 2009). Fracture permeability will also be affected since unfrozen fractures are up to three magnitudes more permeable than frozen ones (Pogrebiskiy and Chernyshev, 1977), with significant effects on hydrostatic pressure. Pressurized water in fractures boosts permafrost degradation and could become more important than thermal propagation.

#### 7 Conclusions

625

630

635

645

650

- This study combines a decade of meteorological data, snowmelt modeling, and discharge measurements, and thereby provides thereby providing (i) novel insights into water dynamics in degraded permafrost bedrock. We describe in detail the timing and the quantities of water infiltration in deep fractures after snowmelt and rainfall events and estimate the possible resulting fractured bedrock and (ii) estimating possible hydrostatic pressures.
  - i) At this elevation, snowmelt produces, on average and in total, higher discharges than rainfall events, while extreme events are similar in both cases and can reach up to 800 l/d. Due to but due to climate change, this dynamic might shift towards more rain and less snowmelt more rain is expected by the end of the century. Extreme events can reach up to 800 l/d from one single fracture. Rainfall reaches the 55m-deep tunnel with an average delay of 31 h, but this value decreases when the fractures are pre-saturated, e.g., during snowmelt periods. No-flow and baseflow events are indicators of unsaturated and partially saturated fractures, respectively. Fully saturated fracturesrequire > To fully saturate the fractures, more than 30 mm precipitation within 24 h is required, but high-intensity short-duration rain barely contributes to fracture flow. The discharge curves of summer precipitation fit into a general recession curve composed of two exponential terms for laminar flow, which allows forecasting the duration of an event, given its discharge. With We detected flow anomalies that can be explained with the help of a simple an empirical fracture flow model, we detected flow anomalies that can be explained by considering saturation changes and water storage. The One fracture can store up to 550 1 in extreme events, which is expected to fully saturate the fracture and increase its interconnectivity.

ii) The hydraulic head resulting from the water accumulation is computed using the recession curve and Darcy's Law for a falling headtogether with the recession curve. On a daily mean, hydrostatic pressures can reach  $27 \text{ m} \pm 6 \text{ m}$  during snowmelt, while rain events generate slightly lower pressures. Snowmelt generates long-lasting pressures for weeks, with daily cycles that can strongly reduce slope stability. Extreme events produce discharges up to 58 l/h in the tunnel, resulting in hydrostatic pressures of  $40 \text{ m} \pm 10 \text{ m}$  (400 kPa). These values are definitely enough to enough to weaken or trigger unstable slopes. Climate change will likely reduce snowmelt pressures and increase summer extreme events.

660

665

Here we quantitatively demonstrate the relevance of water flow in deep fractures and prove its relevance for slope stability of degraded bedrockpermafrost. The estimated hydrostatic pressures can destabilize and/or trigger unstable rock slopes. The combination of climate change and hydrostatic pressures in periglacial areas amplifies permafrost degradation so that in the near future, water is expected to reach new paths and deeper levels, producing higher pressures, thus increasing the hazard.

Code and data availability. Discharge data, modeled snowmelt, precipitation data, and the corresponding codes for data analysis are available at the following online repository https://doi.org/10.5281/zenodo.13833727 (Scandroglio, 2024). Weather data for snow modeling can be obtained from the Bavarian Avalanche Warning Service (Lawinenwarndienst im Bayerischen Landesamt für Umwelt). Weather data from the summit can be obtained from the DWD German Weather Service at https://cdc.dwd.de/portal/.

Author contributions. RS designed the study under the supervision of MK. RS conducted the fieldwork with the support of TR. RS and SW developed the concept of the study, the data analysis, and the model. RS performed the analysis, implemented the model, and made the figures in Matlab. RS prepared the manuscript with revision and final approval from all authors.

Competing interests. At least one of the (co-)authors is a member of the editorial board of Earth Surface Dynamics. The authors have no other competing interests to declare.

Acknowledgements. This study was supported by the AlpSenseRely project, funded by the Bavarian State Ministry of the Environment and Consumer Protection (TUS01UFS-76976), and by the Hydro-PF project, funded by the TUM International Graduate School of Science and Technology IGSSE (Team 12.9). Special thanks to the Environmental Research Station Schneefernerhaus and the Bayerische Zugspitzbahn Bergbahn AG for the amazing logistic support. We thank all the students and colleagues who provided support in more than 100 days of fieldwork. We are also grateful to Franziska and Thomas from the Bavarian Avalanche Center for the snow data.

695

700

- Allan Freeze, R. and Cherry, J. A.: Groundwater, Prentice-Hall, 1979.
- Arenson, L. U., Harrington, J. S., Koenig, C. E. M., and Wainstein, P. A.: Mountain Permafrost Hydrology—A Practical Review Following Studies from the Andes, Geosciences, 12, 48, https://doi.org/10.3390/geosciences12020048, 2022.
- Banks, E. W., Simmons, C. T., Love, A. J., Cranswick, R., Werner, A. D., Bestland, E. A., Wood, M., and Wilson, T.: Fractured bedrock and saprolite hydrogeologic controls on groundwater/surface-water interaction: A conceptual model (Australia), Hydrogeology Journal, 17, 1969–1989, https://doi.org/10.1007/s10040-009-0490-7, 2009.
  - Bast, A., Kenner, R., and Phillips, M.: Short-term cooling, drying and deceleration of an ice-rich rock glacier, EGUSphere Discussion, https://doi.org/10.5194/egusphere-2024-269, 2024.
- Ben-Asher, M., Magnin, F., Westermann, S., Bock, J., Malet, E., Berthet, J., Ravanel, L., and Deline, P.: Estimating surface water availability in high mountain rock slopes using a numerical energy balance model, Earth Surf. Dynam, 11, 899–915, https://doi.org/10.5194/esurf-11-899-2023, 2023.
  - Boussinesq, M. J.: Essai sur la theories des eaux courantes., Memoires presentes par divers savants a l'Academie des Sciences de l'Institut National de France, tome xxiii edn., 1877.
  - Clarke, B. A. and Burbank, D. W.: Quantifying bedrock-fracture patterns within the shallow subsurface: Implications for rock mass strength, bedrock landslides, and erodibility, Journal of Geophysical Research: Earth Surface, 116, https://doi.org/10.1029/2011JF001987, 2011.
  - Clow, D. W., Schrott, L., Webb, R., Campbell, D. H., Torizzo, A., and Dornblaser, M.: Ground Water Occurrence and Contributions to Streamflow in an Alpine Catchment, Colorado Front Range, GroundWater, 41, 937–950, https://doi.org/10.1111/J.1745-6584.2003.TB02436.X, 2003.
  - Cochand, M., Christe, P., Ornstein, P., and Hunkeler, D.: Groundwater Storage in High Alpine Catchments and Its Contribution to Streamflow, Water Resources Research, 55, 2613–2630, https://doi.org/10.1029/2018WR022989, 2019.
    - Courtin, G. M. and Bliss, L. C.: A Hydrostatic Lysimeter to Measure Evapotranspiration under Remote Field Conditions, Arctic and Alpine Research, 3, 81, https://doi.org/10.2307/1550384, 1971.
    - Draebing, D., Krautblatter, M., and Hoffmann, T.: Thermo-cryogenic controls of fracture kinematics in permafrost rockwalls, Geophysical Research Letters, 44, 3535–3544, https://doi.org/10.1002/2016GL072050, 2017.
- 705 Etzelmüller, B., Czekirda, J., Magnin, F., Duvillard, P. A., Ravanel, L., Malet, E., Aspaas, A., Kristensen, L., Skrede, I., Majala, G. D., Jacobs, B., Leinauer, J., Hauck, C., Hilbich, C., Böhme, M., Hermanns, R., Eriksen, H. O., Lauknes, T. R., Krautblatter, M., and Westermann, S.: Permafrost in monitored unstable rock slopes in Norway-New insights from temperature and surface velocity measurements, geophysical surveying, and ground temperature modelling, Earth Surface Dynamics, 10, 97–129, https://doi.org/10.5194/ESURF-10-97-2022, 2022.
  - Fernandez, G. and Moon, J.: Excavation-induced hydraulic conductivity reduction around a tunnel Part 2: Verification of proposed method using numerical modeling, Tunnelling and Underground Space Technology, 25, 567–574, https://doi.org/10.1016/j.tust.2010.04.001, 2010.
  - Fischer, L., Amann, F., Moore, J. R., and Huggel, C.: Assessment of periglacial slope stability for the 1988 Tschierva rock avalanche (Piz Morteratsch, Switzerland), Engineering Geology, 116, 32–43, https://doi.org/10.1016/J.ENGGEO.2010.07.005, 2010.
  - Gabrielli, C., McDonnell, J., and Jarvis, W.: The role of bedrock groundwater in rainfall–runoff response at hillslope and catchment scales, Journal of Hydrology, 450-451, 117–133, https://doi.org/10.1016/j.jhydrol.2012.05.023, 2012.
- Gallemann, T., Haas, U., Teipel, U., von Poschinger, A., Wagner, B., Mahr, M., and Bäse, F.: Permafrost-Messstation am Zugspitzgipfel: Ergebnisse und Modellberechnungen, Tech. rep., https://www.bestellen.bayern.de/shoplink/91115.htm, 2017.

- Gruber, S. and Haeberli, W.: Permafrost in steep bedrock slopes and its temperatures-related destabilization following climate change, Journal of Geophysical Research: Earth Surface, 112, 1–10, https://doi.org/10.1029/2006JF000547, 2007.
- Haeberli, W. and Gruber, S.: Global Warming and Mountain Permafrost, Permafrost Soils, pp. 205–218, https://doi.org/10.1007/978-3-540-720 69371-0\_14, 2009.
  - Haeberli, W., Wegmann, M., and Vonder Muehll, D.: Slope stability problems related to glacier shrinkage and permafrost degradation in the Alps, Eclogae geol. Helv., 90, 407–414, https://doi.org/https://doi.org/10.5169/seals-168172, 1997.
  - Hasler, A., Gruber, S., and Beutel, J.: Kinematics of steep bedrock permafrost, Journal of Geophysical Research: Earth Surface, 117, 1–17, https://doi.org/10.1029/2011JF001981, 2012.
- Hauck, C., Böttcher, M., and Maurer, H.: A new model for estimating subsurface ice content based on combined electrical and seismic data sets, The Cryosphere, 5, 453–468, https://doi.org/10.5194/tc-5-453-2011, 2011.
  - Hayashi, M.: Alpine Hydrogeology: The Critical Role of Groundwater in Sourcing the Headwaters of the World, Groundwater, 58, 498–510, https://doi.org/10.1111/gwat.12965, 2020.
- Helmstetter, A. and Garambois, S.: Seismic monitoring of Séchilienne rockslide (French Alps): Analysis of seismic signals and their correlation with rainfalls, Journal of Geophysical Research: Earth Surface, 115, https://doi.org/10.1029/2009JF001532, 2010.
  - Hornung, T. and Haas, U.: Erlaeuterungen zu den Blaettern 8531/8631 Zugspitze und 8531/8632 Garmisch-Partenkirchen, Tech. rep., Bayrisches Landesamt für Umwelt, Augsburg, 2017.
  - Immerzeel, W. W., Lutz, A. F., Andrade, M., Bahl, A., Biemans, H., Bolch, T., Hyde, S., Brumby, S., Davies, B. J., Elmore, A. C., Emmer, A., Feng, M., Fernández, A., Haritashya, U., Kargel, J. S., Koppes, M., Kraaijenbrink, P. D. A., Kulkarni, A. V., Mayewski, P. A., Nepal,
- S., Pacheco, P., Painter, T. H., Pellicciotti, F., Rajaram, H., Rupper, S., Sinisalo, A., Shrestha, A. B., Viviroli, D., Wada, Y., Xiao, C., Yao, T., and Baillie, J. E. M.: Importance and vulnerability of the world's water towers, Nature, 577, 364–369, https://doi.org/10.1038/s41586-019-1822-y, 2020.
  - IPCC: Weather and Climate Extreme Events in a Changing Climate, in: Climate Change 2021 The Physical Science Basis, pp. 1513–1766, Cambridge University Press, https://doi.org/10.1017/9781009157896.013, 2023.
- Jones, D., Harrison, S., Anderson, K., and Betts, R. A.: Mountain rock glaciers contain globally significant water stores, Scientific Reports, 8, 2834, https://doi.org/10.1038/s41598-018-21244-w, 2018.
  - Kirchner, J. W.: Catchments as simple dynamical systems: Catchment characterization, rainfall-runoff modeling, and doing hydrology backward, Water Resources Research, 45, https://doi.org/10.1029/2008WR006912, 2009.
- Kneisel, C., Rödder, T., and Schwindt, D.: Frozen ground dynamics resolved by multi-year and yearround electrical resistivity monitoring at three alpine sites in the Swiss Alps, Near Surface Geophysics, 12, 117–132, https://doi.org/10.3997/1873-0604.2013067, 2014.
  - Krautblatter, M.: Detection and quantification of permafrost change in alpine rock walls and implications for rock instability, 2009.
  - Krautblatter, M., Verleysdonk, S., Flores-Orozco, A., and Kemna, A.: Temperature-calibrated imaging of seasonal changes in permafrost rock walls by quantitative electrical resistivity tomography (Zugspitze, German/Austrian Alps), Journal of Geophysical Research: Earth Surface, 115, 1–15, https://doi.org/10.1029/2008JF001209, 2010.
- Krautblatter, M., Funk, D., and Günzel, F. K.: Why permafrost rocks become unstable: a rock-ice-mechanical model in time and space, Earth Surface Processes and Landforms, 38, 876–887, https://doi.org/10.1002/esp.3374, 2013.
  - Kristensen, L., Czekirda, J., Penna, I., Etzelmüller, B., Nicolet, P., Pullarello, J. S., Blikra, L. H., Skrede, I., Oldani, S., and Abellan, A.: Movements, failure and climatic control of the Veslemannen rockslide, Western Norway, Landslides, 18, 1963–1980, https://doi.org/10.1007/s10346-020-01609-x, 2021.

- Lehning, M., Bartelt, P., Brown, B., Russi, T., Stöckli, U., and Zimmerli, M.: SNOWPACK model calculations for avalanche warning based upon a new network of weather and snow stations, Cold Regions Science and Technology, 30, 145–157, https://doi.org/10.1016/S0165-232X(99)00022-1, 1999.
  - Leinauer, J., Dietze, M., Knapp, S., Scandroglio, R., Jokel, M., and Krautblatter, M.: How water, temperature, and seismicity control the preconditioning of massive rock slope failure (Hochvogel), Earth Surf. Dynam, 12, 1027–1048, https://doi.org/10.5194/esurf-12-1027-2024, 2024.
  - Levy, W., Pandelova, M., Henkelmann, B., Bernhöft, S., Fischer, N., Antritter, F., and Schramm, K. W.: Persistent organic pollutants in shallow percolated water of the Alps Karst system (Zugspitze summit, Germany), Science of the Total Environment, 579, 1269–1281, https://doi.org/10.1016/j.scitotenv.2016.11.113, 2017.
- Magnin, F. and Josnin, J.: Water Flows in Rock Wall Permafrost: A Numerical Approach Coupling Hydrological and Thermal Processes,

  Journal of Geophysical Research: Earth Surface, 126, https://doi.org/10.1029/2021JF006394, 2021.
  - Maillet, E.: Essais d'Hydraulique souterraine et fluviale, Hermann, Paris, 1905.

- Malík, P.: Evaluating Discharge Regimes of Karst Aquifer, in: Karst Aquifers—Characterization and Engineering. Professional Practice in Earth Sciences., edited by Stevanović, Z., chap. 7, pp. 205–249, Springer, Cham, https://doi.org/10.1007/978-3-319-12850-4\_7, 2015.
- Malík, P. and Vojtková, S.: Use of recession-curve analysis for estimation of karstification degree and its application in assessing overflow/underflow conditions in closely spaced karstic springs, Environmental Earth Sciences, 65, 2245–2257, https://doi.org/10.1007/S12665-012-1596-0/TABLES/4, 2012.
  - Mamot, P., Weber, S., Eppinger, S., and Krautblatter, M.: A temperature-dependent mechanical model to assess the stability of degrading permafrost rock slopes, Earth Surface Dynamics, 9, 1125–1151, https://doi.org/10.5194/esurf-9-1125-2021, 2021.
- Manning, A. and Caine, J.: Groundwater noble gas, age, and temperature signatures in an Alpine watershed: Valuable tools in conceptual model development, Water Resources Research, 43, https://doi.org/10.1029/2006WR005349, 2007.
  - Markovich, K. H., Manning, A. H., Condon, L. E., and McIntosh, J. C.: Mountain-Block Recharge: A Review of Current Understanding, Water Resources Research, 55, 8278–8304, https://doi.org/10.1029/2019WR025676, 2019.
  - Mayer, C., Weber, M., Wendt, A., and Hagg, W.: Die bayerischen Gletscher, die verbliebenen Eisreserven Deutschlands, Polarforschung, 89, 1–7, https://doi.org/10.5194/POLF-89-1-2021, 2021.
- Montgomery, D., Dietrich, W., and Heffner, J.: Piezometric response in shallow bedrock at CB1: Implications for runoff generation and landsliding, Water Resources Research, 38, 10–1–10–18, https://doi.org/10.1029/2002WR001429, 2002.
  - Morche, D., Witzsche, M., and Schmidt, K. H.: Hydrogeomorphological characteristics and fluvial sediment transport of a high mountain river (Reintal Valley, Bavarian Alps, Germany), Zeitschrift für Geomorphologie, Supplementary Issues, 52, 51–77, https://doi.org/10.1127/0372-8854/2008/0052S1-0051, 2008.
- Musselman, K. N., Clark, M. P., Liu, C., Ikeda, K., and Rasmussen, R.: Slower snowmelt in a warmer world, Nature Climate Change, 7, 214–219, https://doi.org/10.1038/nclimate3225, 2017.
  - Noetzli, J. and Phillips, M.: Mountain Permafrost Hydrology, Tech. rep., WSL Institute for Snow and Avalanche Research SLF, Davos, Switzerland, https://doi.org/10.16904/slf.1, 2019.
- Pavoni, M., Boaga, J., Wagner, F., Bast, A., and Phillips, M.: Characterization of rock glaciers environments combining structurally-coupled and petrophysically-coupled joint inversions of electrical resistivity and seismic refraction datasets, Journal of Applied Geophysics, 215, 105 097, https://doi.org/10.1016/J.JAPPGEO.2023.105097, 2023.

- Phillips, M., Haberkorn, A., Draebing, D., Krautblatter, M., Rhyner, H., and Kenner, R.: Seasonally intermittent water flow through deep fractures in an Alpine Rock Ridge: Gemsstock, Central Swiss Alps, Cold Regions Science and Technology, 125, 117–127, https://doi.org/10.1016/j.coldregions.2016.02.010, 2016.
- Phillips, M., Wolter, A., Lüthi, R., Amann, F., Kenner, R., and Bühler, Y.: Rock slope failure in a recently deglaciated permafrost rock wall at Piz Kesch (Eastern Swiss Alps), February 2014, Earth Surface Processes and Landforms, 42, 426–438, https://doi.org/10.1002/esp.3992, 2017.

800

- Phillips, M., Buchli, C., Weber, S., Boaga, J., Pavoni, M., and Bast, A.: Brief communication: Combining borehole temperature, borehole piezometer and cross-borehole electrical resistivity tomography measurements to investigate seasonal changes in ice-rich mountain permafrost, The Cryosphere, 17, 753–760, https://doi.org/10.5194/tc-17-753-2023, 2023.
- Pogrebiskiy, M. I. and Chernyshev, S. N.: Determination of the permeability of the frozen fissured rock massif in the vicinity of the Kolyma hydroelectric power station, Cold Regions Research and Engineering Laboratory Draft translation, 634, 1–13, 1977.
- POPALP Report: Erfassung von persistenten organischen Schadstoffen im bayerischen Alpenraum, Tech. rep., München, Helmholtz Zentrum, https://www.lfu.bayern.de/luft/schadstoffe\_luft/projekte/doc/schlussbericht\_popalp\_quellwasser.pdf, 2011.
- Poulain, A., Watlet, A., Kaufmann, O., Van Camp, M., Jourde, H., Mazzilli, N., Rochez, G., Deleu, R., Quinif, Y., and Hallet, V.: Assessment of groundwater recharge processes through karst vadose zone by cave percolation monitoring, Hydrological Processes, 32, 2069–2083, https://doi.org/10.1002/HYP.13138, 2018.
  - Rappl, A., Wetzel, K., Büttner, G., and Scholz, M.: Tracerhydrologische Untersuchungen am Partnach-Ursprung, Hydrologie und Wasser bewirtschaftung, 4, 220–230, http://www.hywa-online.de/landnutzungswandel-und-wasserressourcen-im-bundesdistrikt-zentral-brasilien/, 2010.
  - Rist, A. and Phillips, M.: First results of investigations on hydrothermal processes within the active layer above alpine permafrost in steep terrain, Norsk Geografisk Tidsskrift Norwegian Journal of Geography, 59, 177–183, https://doi.org/10.1080/00291950510020574, 2005.
  - Roth, M. and Blikra, L. H.: Seismic monitoring of the unstable rock slope at Aaknes, Norway, Geophysical Research Abstracts, 11, https://meetingorganizer.copernicus.org/EGU2009/EGU2009-3680.pdf, 2009.
- 815 Scandroglio, R.: Collection of data and Matlab-codes for analyzing water dynamic the Zugspitze., at https://doi.org/10.5281/zenodo.13833727, 2024.
  - Scandroglio, R., Stoll, V., and Krautblatter, M.: The driving force of all nature. Modelling water pressure and its stability consequences on alpine bedrock slopes, IOP Conference Series: Earth and Environmental Science, 833, 012 109, https://doi.org/10.1088/1755-1315/833/1/012109, 2021.
- Stoffel, M. and Huggel, C.: Effects of climate change on mass movements in mountain environments, Progress in Physical Geography, 36, 421–439, https://doi.org/10.1177/0309133312441010/ASSET/IMAGES/LARGE/10.1177\_0309133312441010-FIG7.JPEG, 2012.
  - Sweetenham, M. G., Maxwell, R. M., and Santi, P. M.: Assessing the timing and magnitude of precipitation-induced seepage into tunnels bored through fractured rock, Tunnelling and Underground Space Technology, 65, 62–75, https://doi.org/10.1016/j.tust.2017.02.003, 2017.
- Ulrich, R. and King, L.: Influence of mountain permafrost on construction in the Zugspitze mountains, Bavarian alps, Germany, in: International conference; 6th, Permafrost; 1993; Beijing, pp. 625–630, South China University of Technology Press, https://www.tib.eu/en/search/id/BLCP%3ACN015187124/Influence-of-Mountain-Permafrost-on-Construction/, 1993.
  - van Tiel, M., Aubry-Wake, C., Somers, L., Andermann, C., Avanzi, F., Baraer, M., Chiogna, G., Daigre, C., Das, S., Drenkhan, F., Farinotti, D., Fyffe, C. L., de Graaf, I., Hanus, S., Immerzeel, W., Koch, F., McKenzie, J. M., Müller, T., Popp, A. L., Saidaliyeva, Z., Schaefli, B.,

- Schilling, O. S., Teagai, K., Thornton, J. M., and Yapiyev, V.: Cryosphere–groundwater connectivity is a missing link in the mountain water cycle, Nature Water, 2, 624–637, https://doi.org/10.1038/s44221-024-00277-8, 2024.
  - Viviroli, D., Dürr, H. H., Messerli, B., Meybeck, M., and Weingartner, R.: Mountains of the world, water towers for humanity: Typology, mapping, and global significance, 43, 7447, https://doi.org/10.1029/2006WR005653, 2007.
- Voigt, C., Schulz, K., Koch, F., Wetzel, K.-F., Timmen, L., Rehm, T., Pflug, H., Stolarczuk, N., Förste, C., and Flechtner, F.: Technical note: Introduction of a superconducting gravimeter as novel hydrological sensor for the Alpine research catchment Zugspitze, Hydrology and Earth System Sciences, 25, 5047–5064, https://doi.org/10.5194/HESS-25-5047-2021, 2021.
  - Vorkauf, M., Marty, C., Kahmen, A., and Hiltbrunner, E.: Past and future snowmelt trends in the Swiss Alps: the role of temperature and snowpack, Climatic Change, 165, 1–19, https://doi.org/10.1007/S10584-021-03027-X/TABLES/3, 2021.
  - Wagner, F., Mollaret, C., Günther, T., Kemna, A., and Hauck, C.: Quantitative imaging of water, ice and air in permafrost systems through petrophysical joint inversion of seismic refraction and electrical resistivity data, Geophysical Journal International, 219, 1866–1875, https://doi.org/10.1093/gji/ggz402, 2019.
  - Walter, F., Amann, F., Kos, A., Kenner, R., Phillips, M., de Preux, A., Huss, M., Tognacca, C., Clinton, J., Diehl, T., and Bonanomi, Y.: Direct observations of a three million cubic meter rock-slope collapse with almost immediate initiation of ensuing debris flows, Geomorphology, 351, 106 933, https://doi.org/10.1016/j.geomorph.2019.106933, 2020.
- Walvoord, M. A. and Kurylyk, B. L.: Hydrologic Impacts of Thawing Permafrost—A Review, Vadose Zone Journal, 15, 1–20, https://doi.org/10.2136/vzj2016.01.0010, 2016.
  - Watlet, A., Kaufmann, O., Triantafyllou, A., Poulain, A., Chambers, J. E., Meldrum, P. I., Wilkinson, P. B., Hallet, V., Quinif, Y., Van Ruymbeke, M., and Camp, M. V.: Imaging groundwater infiltration dynamics in the karst vadose zone with long-term ERT monitoring, Hydrol. Earth Syst. Sci, 22, 1563–1592, https://doi.org/10.5194/hess-22-1563-2018, 2018.
- Weber, M., Bernhardt, M., Pomeroy, J. W., Fang, X., Härer, S., and Schulz, K.: Description of current and future snow processes in a small basin in the Bavarian Alps, Environmental Earth Sciences, 75, 1–18, https://doi.org/10.1007/S12665-016-6027-1/FIGURES/9, 2016.
  - Weber, M., Koch, F., Bernhardt, M., and Schulz, K.: The evaluation of the potential of global data products for snow hydrological modelling in ungauged high-alpine catchments, Hydrol. Earth Syst. Sci, 25, 2869–2894, https://doi.org/10.5194/hess-25-2869-2021, 2021.
  - Weber, S., Beutel, J., Faillettaz, J., Hasler, A., Krautblatter, M., and Vieli, A.: Quantifying irreversible movement in steep, fractured bedrock permafrost on Matterhorn (CH), The Cryosphere, 11, 567–583, https://doi.org/10.5194/tc-11-567-2017, 2017.
- 855 Weishaupt, S.: Hochgebirgshydrologie, 2021.

840

- Welch, L. A. and Allen, D. M.: Hydraulic conductivity characteristics in mountains and implications for conceptualizing bedrock groundwater flow, Hydrogeology Journal, 22, 1003–1026, https://doi.org/10.1007/s10040-014-1121-5, 2014.
- Wetzel, K.: On the hydrology of the Partnach area in the Wetterstein mountains (Bavarian Alps), Erdkunde, 58, 172–186, 2004.
- Wetzel, K., Bernhardt, M., Weishaupt, S., and Weber, M.: Hydrological investigations in the Wetterstein Mountains at the UFS Schneefernerhaus (Bavarian Alps), in: Science at the Environmental Research Station Schneefernerhaus / Zugspitze, edited by Bittner, M., chap. 19, pp. 305–321, 2022.
  - Woo, M. K.: Permafrost hydrology, vol. 9783642234, https://doi.org/10.1007/978-3-642-23462-0, 2012.
  - Zha, Y., Yang, J., Zeng, J., Tso, C. H. M., Zeng, W., and Shi, L.: Review of numerical solution of Richardson–Richards equation for variably saturated flow in soils, Wiley Interdisciplinary Reviews: Water, 6, https://doi.org/10.1002/WAT2.1364, 2019.

865 Zhou, Q., Salve, R., Liu, H.-H., Wang, J. S., and Hudson, D.: Analysis of a mesoscale infiltration and water seepage test in unsaturated fractured rock: Spatial variabilities and discrete fracture patterns, Journal of Contaminant Hydrology, 87, 96–122, https://doi.org/10.1016/j.jconhyd.2006.05.001, 2006.

## **Supplementary material**

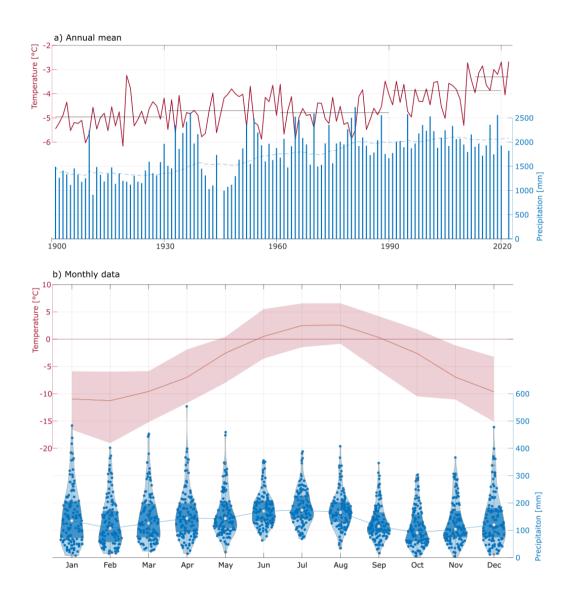
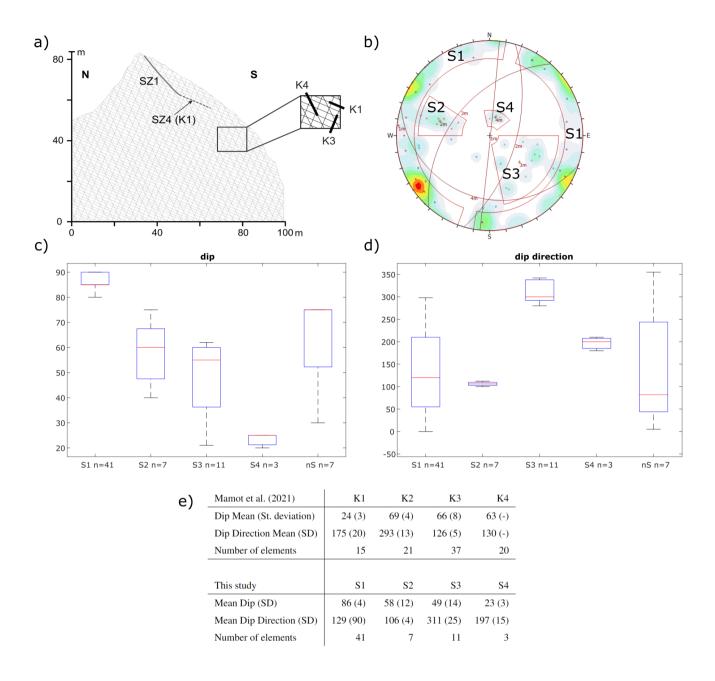


Figure S1. Long-term and seasonal temperature and precipitation patterns.

- a) The red line represents the annual mean temperature, and the black lines are the respective averages of the selected period. It can be seen that the last 3 years have a drastical increase, and the last 10 years even more. Blue bars represent the annual sum of precipitation, the blue line is the 30-year moving mean: no recent changes are visible.
- b) Monthly data between 1901 and 2022. The average temperature is shown in red, with  $\pm 2 \cdot standarddeviation$ . The blue violin plots represent the sum of precipitation for each month using a kernel distribution that highlights the 1st and 3rd quantiles with darker blue and the median with a white point. Summer maximum, February and October minimum.



**Figure S2.** Fracture mapping. a) NS-Profile of the Zugspitze Ridge with the instability analyzed analyzed in Mamot et al. (2021), with fractures from the external scan line, which is located 400m from the tunnel. b) Results from the underground scan line in the tunnel, presented in this study. c) and d) Dip and dip direction for the 5 clusters of the tunnel scan line in the tunnel. e) Numerical comparison of the two scan lines.

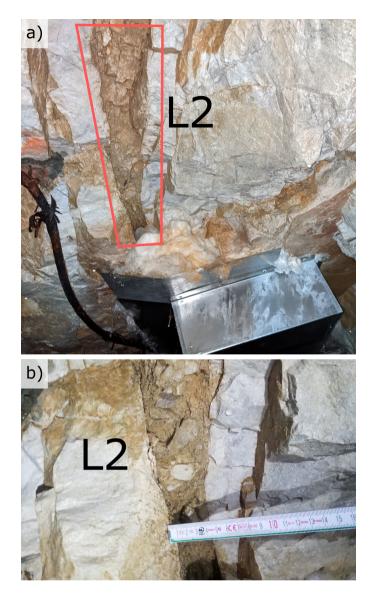


Figure S3. a) Picture of fracture L2. b) Measurement of fracture aperture.

Table S1. Modelling Parameters of all discharge events with a uniform recession curve. a) All events are plotted with  $t_0 = 0$ . Semi-logarithmic axes are used to improve visualisation. b) Discharge curves are shifted in time so that the value of  $Q_0$  at  $t_0$  fits a similar value from a higher curve. In blue the mean recession curve .e) Recession curves with single equation linear flow component (green line) and single equation exponential flow (blue line), trying to fit the mean curve. While the linear solution cannot reproduce the storage curve, the long-term exponential component fits the values well until 20, but not smaller discharges. d) Recession curve with double-component exponential flow, fittingwell the mean curve.

Fitting	Flow component	$\alpha_x$ (or $\beta_x$ ) [1/h]	$Q0_x[1/h]$	Error <sup>1</sup>
Linear	fast-flow / turbulent	$\beta_1 = 0.023$	$Q0_1 = 13.6$	<u>4103 1</u>
Exponential 1	slow-flow / laminar	$\alpha_1 = 0.109$	$Q0_1 = 41.9$	$\underbrace{861}$
Exponential 2	slow-flow / laminar	$\alpha_1 = 0.015 / \alpha_2 = 0.123$	$Q0_1 = 1.8 / Q0_1 = 41.4$	17.1

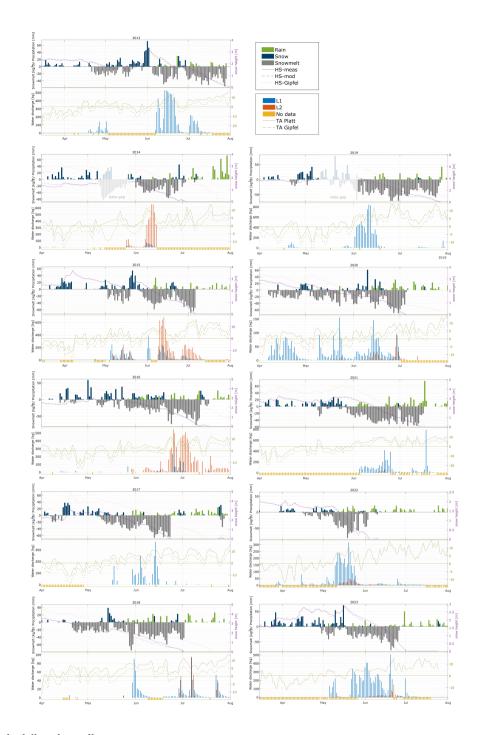
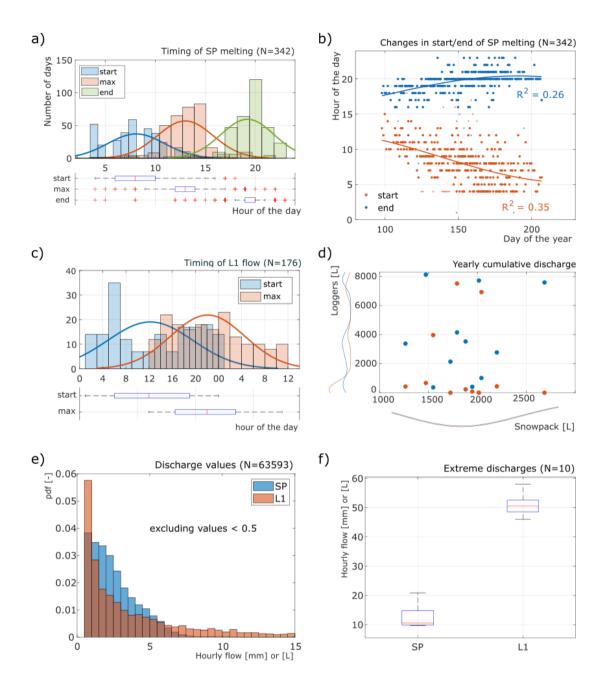
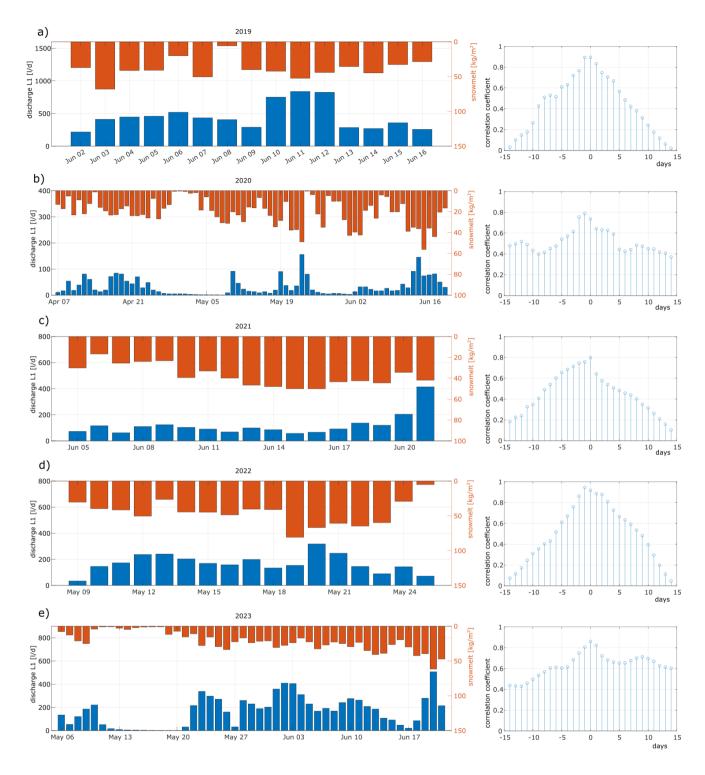


Figure S4. Snowmelt, daily values, all years.



**Figure S5.** Additional statistics. a) Timing of start, peak flow, and end of snowmelt, with probability distributions and boxplots. b) Variability in the start and the end of melting during the season, fitting quite well with a sinus curve. High timing variability is due to daily different sun radiation, cloud cover and temperature. c) Timing of start and max flow in L1, with a probability distribution and boxplot. d) Sum of water flow for each year. e) Distribution of discharge values for the model, in blue, and for L1, in red. In both cases, there is a very high concentration of small values (<0.5 L or mm) that has been excluded. g) Ten maximum hourly discharge events for the model and for L1.



**Figure S6.** Daily correlation between snowmelt and water flow, from 2019 (a) to 2023 (e). On the left: Snowmelt in red and water flow in L1 in blue. On the right: results of the correlation for each single year.

	START TIME	TIME				END TIME		٦٥	DURATION			_	TOTAL			MA	MAXIMUM		2	MAX 24	
PSUM	11	12	Delay R-L1 [h]	Delay R-L2 [h]	PSUM	11	71	PSUM [h]	11 E	- 7 [h]	PSUM [mm]	11 E		L1/ PSUM P	PSUM [r	PSUM [mm/h]	[-, I	[1/J]	PSUM [mm/d]	11 [[/d]	[1/d]
11/8/14 - 02	12/8/14 - 00	11/8/14 - 16	22	14	14/8/14 - 15	20/8/14 - 11	20/8/14 - 17	82	203	217	87	190	1338	2.2	15.4	6.8	5.3	15.9	47.9	75.5	347.5
15/8/15 - 13	16/8/15 - 20	16/8/15 - 20	31	31	19/8/15 - 11	22/8/15 - 21	24/8/15 - 07	75	145	179	104	464	1945	4.4	18.7	6.3	12.5	24.6	43.3	197.1	570.8
27/6/18 - 12	29/6/18 - 14	29/6/18 - 16	20	52	28/6/18 - 23	02/7/18 - 05	01/7/18 - 11	35	63	43	49	81	99	1.7	1.3	4.1	2.5	2.8	39.1	50.9	53.7
05/7/18 - 11	06/7/18 - 20	06/7/18 - 20	33	33	07/7/18 - 07	10/7/18 - 04	09/7/18 - 04	44	80	95	71	170	155	2.4	2.2	8.1	6.4	5.1	52.7	101.8	104.0
22/7/18 - 21	23/7/18 - 12	23/7/18 - 15	15	18	23/7/18 - 13	26/7/18 - 00	25/7/18 - 00	16	09	33	33	108	53	3.3	1.6	13.9	5.9	2.7	35.2	80.7	49.0
30/8/18 - 03	31/8/18 - 03	31/8/18 - 06	24	27	01/9/18 - 18	04/9/18 - 15	04/9/18 - 14	63	108	104	81	280	233	3.4	2.9	5.4	8.0	4.1	54.3	116.4	79.5
12/7/19 - 01	13/7/19 - 15	-	38		16/7/19 - 00	21/7/19 - 10	-	95	187	•	65	108	•	1.7	•	9.9	2.3	'	44.7	37.0	•
11/8/19 - 20	13/8/19 - 03	13/8/19 - 15	31	43	14/8/19 - 04	16/8/19 - 04	14/8/19 - 11	95	73	20	69	109	25	1.6	0.4	11.3	14.7	7.7	55.2	72.6	25.1
20/8/19 - 14	21/8/19 - 20		30		21/8/19 - 04	23/8/19 - 18		14	46	•	35	45	'	1.3	•	5.2	1.7	'	37.1	31.6	'
01/9/19 - 15	02/9/19 - 07	02/9/19 - 14	16	23	02/9/19 - 21	05/9/19 - 09	03/9/19 - 23	30	74	33	53	118	35	2.2	9.0	11.7	4.2	1.7	51.7	75.3	31.8
28/6/20 - 16	29/6/20 - 11	29/6/20 - 08	19	16	29/6/20 - 15	30/6/20 - 21	30/6/20 - 15	23	34	31	46	133	109	5.9	2.4	6.0	10.0	6.4	46.1	118.9	104.4
10/7/20 - 14		12/7/20 - 03		37	11/7/20 - 19		13/7/20 - 01	23	•	22	26		15	•	0.3	8.0	•	1.5	53.8		15.3
28/8/20 - 17	30/8/20 - 00	-	31		30/8/20 - 15	31/8/20 - 03		46	27	•	65	•	•	•	•	7.6	3.1	'	51.5	58.0	•
08/7/21 - 10	09/7/21 - 05		19		11/7/21 - 06	11/7/21 - 14		89	57	•	78	148	•	1.9	•	10.3	8.2	'	48.6	101.0	
10/7/21 - 16	11/7/21 - 13		21		11/7/21 - 06	14/7/21 - 01		14	09	- 1	29	72	•	2.5		9.2	2.6	'	28.8	46.2	
16/7/21 - 18	71 - 12/2/1		23		19/7/21 - 08	20/7/21 - 01		29	99	•	195	996	•	4.9	•	18.9	55.1	'	157.9	805.3	
03/8/21 - 13	05/8/21 - 12	05/8/21 - 15	47	20	06/8/21 - 08	08/8/21 - 18	07/8/21 - 19	<i>L</i> 9	78	52	62	140	28	2.3	0.5	6.9	4.6	1.2	36.7	81.0	19.5
07/8/21 - 15	08/8/21 - 20	07/8/21 - 15	29		08/8/21 - 19	11/8/21 - 01	09/8/21 - 10	28	23	43	35	49	9	1.4	0.2	9.4	1.6	0.2	34.1	31.4	8.1
22/8/21 - 03	23/8/21 - 13		34		24/8/21 - 15	29/8/21 - 22		09	153	•	95	63	•	1.1	•	9.6	1.4	'	40.5	25.2	
15/9/21 - 13	17/9/21 - 14	-	49		17/9/21 - 05	19/9/21 - 20		40	54	•	51	41	•	0.8	•	9.8	1.5	•	37.7	27.4	•
19/9/21 - 12	20/9/21 - 22	-	34		20/9/21 - 13	26/9/21 - 07		25	129	•	44	86	•	2.3	•	4.4	3.0		43.7	50.5	
26/9/21 - 15	28/9/21 - 13		46		26/9/21 - 23	30/9/21 - 20		∞	55		24	18		0.8		9.0	0.5	'	24.0	10.3	-
23/6/22 - 20	25/6/22 - 06	'	34	•	24/6/22 - 20	27/6/22 - 15		24	57	-	33	80	•	2.4		11.0	3.6	-	42.0	56.9	-

Figure S7. List of selected summer precipitation events with water flow in tunnel. All dates are expressed with the format "dd/m/yy - hh".

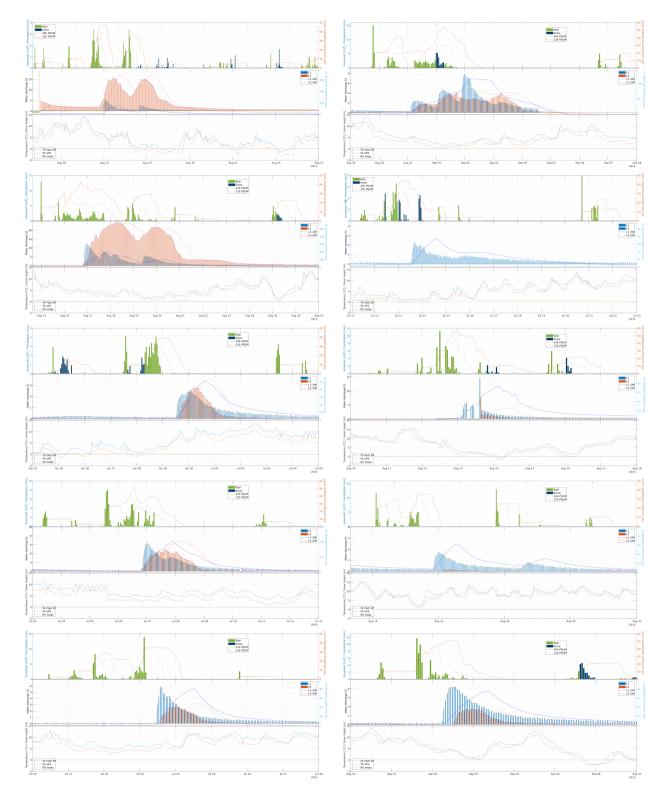


Figure S8. Summer precipitation events, hourly values - part 1

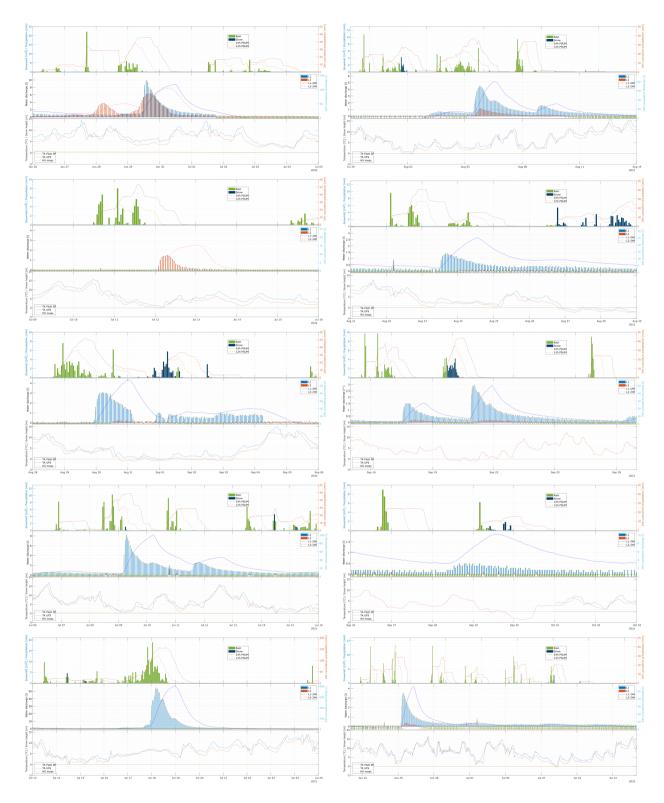
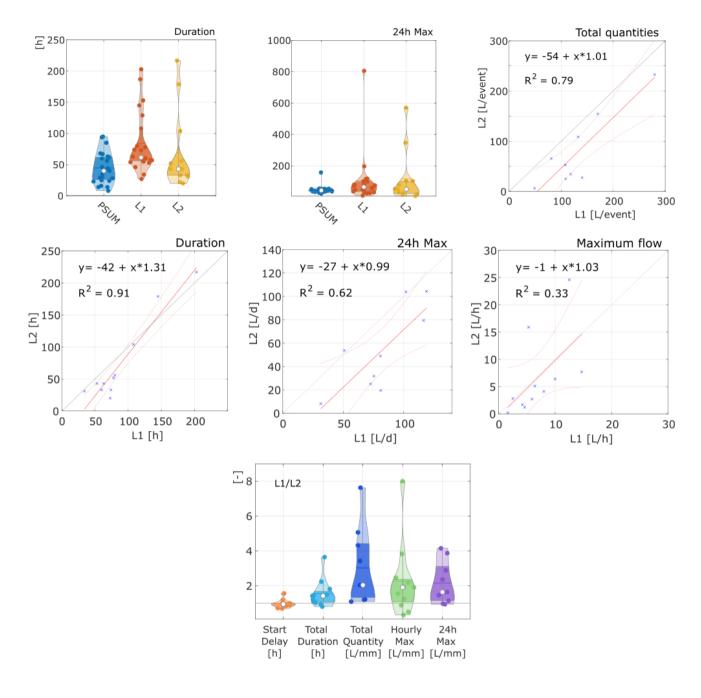


Figure S9. Summer precipitation events, hourly values - part 2



**Figure S10.** Violin plots for quantities (upper graphs) and ratios (lower graphs) for L1, L2, and PSUM P for the parameters duration, total quantity, maximum hourly flow, and maximum daily flow.

Start event	Stop event	Duration (h)	Total precipitation (mm)	Maximum precipitation (mm/h)
8/9/2014 3:00	8/9/2014 15:00	12:00:00	10.80	3.90
9/7/201416:00	9/7/2014 17:00	01:00:00	1.70	1.60
9/8/201417:00	9/8/2014 19:00	02:00:00	8.02	6.95
8/24/2015 22:00	8/25/2015 10:00	12:00:00	22.32	4.79
8/27/2015 7:00	8/27/2015 10:00	03:00:00	2.30	1.74
8/30/201517:00	8/30/2015 20:00	03:00:00	4.20	3.10
9/13/2015 20:00	9/14/2015 18:00	22:00:00	31.10	7.96
9/16/2015 1:00	9/16/2015 8:00	07:00:00	6.70	2.00
9/17/2015 16:00	9/17/2015 20:00	04:00:00	13.80	4.97
7/21/2016 8:00	7/21/2016 10:00	02:00:00	5.90	3.62
7/22/201616:00	7/22/2016 17:00	01:00:00	10.00	9.70
7/24/201618:00	7/24/2016 23:00	05:00:00	25.22	10.45
7/26/201614:00	7/26/2016 20:00	06:00:00	14.16	11.70
7/28/2016 19:00	7/29/2016 6:00	11:00:00	10.48	3.51
8/2/2016 20:00	8/3/2016 6:00	10:00:00	4.30	1.34
8/15/201610:00	8/15/2016 12:00	02:00:00	5.14	4.44
8/18/2016 1:00	8/18/2016 2:00	01:00:00	3.04	2.82
8/18/2016 19:00	8/19/2016 2:00	07:00:00	4.05	1.65
				4.51
7/28/2018 22:00	7/29/2018 2:00	04:00:00	11.79 2.86	2.38
8/1/201819:00	8/1/2018 21:00	02:00:00	4.21	2.14
8/2/201819:00	8/2/2018 21:00 8/12/2018 21:00			
8/12/201819:00		02:00:00	4.00 34.19	3.96
8/13/201812:00	8/15/2018 0:00	36:00:00		9.55
8/18/201814:00	8/18/2018 22:00	08:00:00	9.58	6.80
8/19/201813:00	8/19/2018 16:00	03:00:00	9.15	4.62
8/21/201815:00	8/21/2018 16:00	01:00:00	0.78	0.78
8/23/201812:00	8/23/2018 21:00	09:00:00	32.17	15.58
7/20/201917:00	7/21/2019 10:00	17:00:00	23.11	10.13
7/27/201912:00	7/27/2019 14:00	02:00:00	24.06	14.75
8/24/201915:00	8/24/2019 18:00	03:00:00	24.76	13.13
8/28/201917:00	8/29/2019 0:00	07:00:00	9.98	3.23
8/31/2019 16:00	8/31/2019 20:00	04:00:00	10.38	4.75
7/13/202111:00	7/13/2021 19:00	08:00:00	18.97	9.76
7/24/202113:00	7/24/2021 19:00	06:00:00	10.00	8.01
7/27/202123:00	7/28/2021 15:00	16:00:00	8.88	3.45
7/30/202114:00	7/31/2021 1:00	11:00:00	18.67	10.80
9/9/202115:00	9/10/2021 0:00	09:00:00	3.09	2.80
9/10/202122:00	9/11/2021 2:00	04:00:00	22.37	7.96
6/5/202213:00	6/5/2022 17:00	04:00:00	8.24	3.54
6/12/202223:00	6/13/2022 10:00	11:00:00	14.13	2.66
6/16/2022 8:00	6/16/2022 11:00	03:00:00	11.22	6.11
6/21/202218:00	6/21/2022 23:00	05:00:00	3.76	2.21
6/22/202216:00	6/22/2022 17:00		12.93	12.93
6/23/202218:00		04:00:00	14.68	6.99
6/27/202218:00	6/27/2022 22:00	04:00:00	18.04	13.19
6/28/202212:00	6/28/2022 20:00		11.73	3.83
6/30/202214:00	7/1/2022 14:00	24:00:00	17.38	6.07
7/3/202223:00	7/5/2022 9:00	34:00:00	27.35	9.52
7/7/202219:00	7/8/2022 3:00	08:00:00	19.37	5.88

**Table S2.** List of selected summer precipitation events with NO water flow in tunnel.

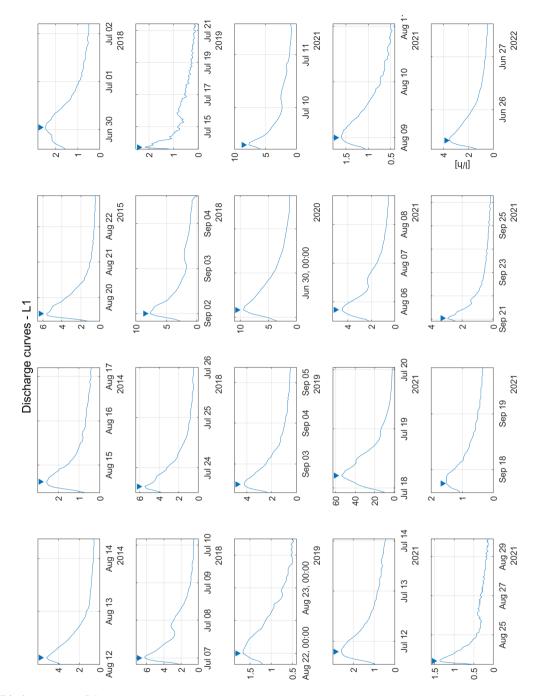
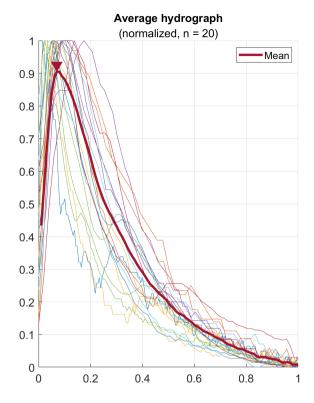
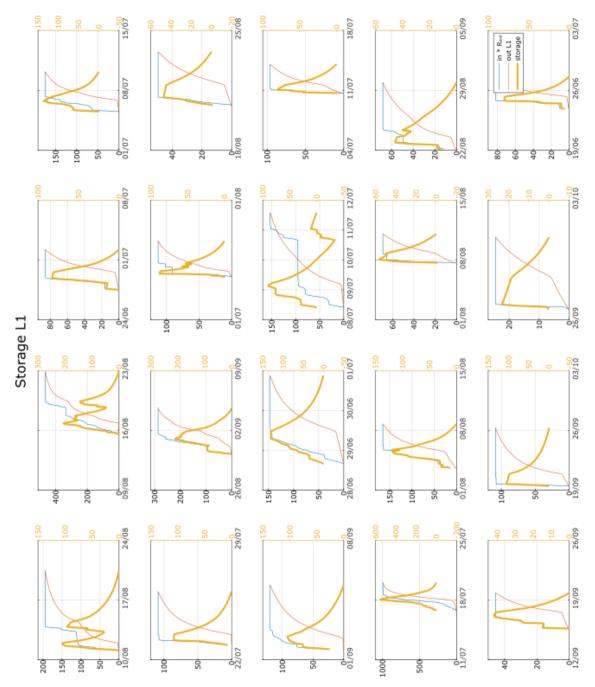


Figure S11. Discharge curves L1



**Figure S12.** Normalized discharge curves for gauge L1, the mean value of the <del>20</del> curves is shown in red.



**Figure S13.** Storage curves for L1. Each graph represents one event: in blue, the cumulative input (precipitation), in red, the cumulative output (gauge L1 or L2), and in yellow, the water that can accumulate into the rock at each hour. The total inflow in every event is forced to be equal to the total outflow, with the help of the multiplication factor R.

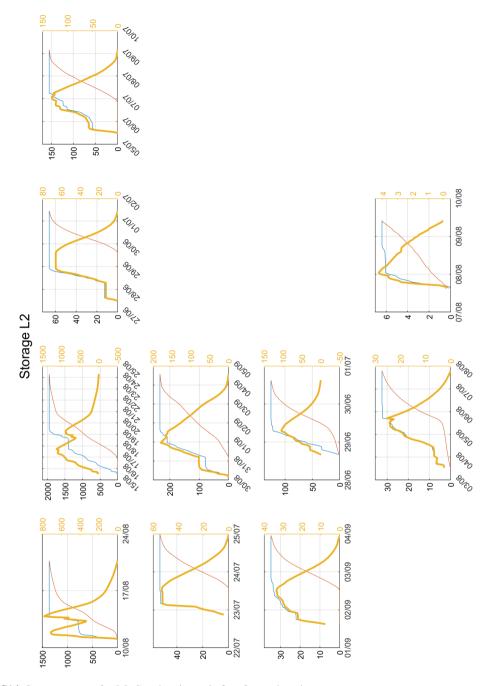


Figure S14. Storage curves for L2. See the picture before for explanations.

K [m/s]	L [m]	t [h]	H1 [m]
1.E-05	1.1	200	69.61
1.E-05	1.2	200	40.34
1.E-05	1.3	200	25.43
1.E-05	1.4	200	17.12
1.E-05	1.5	200	12.15
5.E-05	5.5	200	69.61
5.E-05	5.75	200	52.37
5.E-05	6.1	200	36.56
5.E-05	6.5	200	25.43
5.E-05	7.5	200	12.15
1.E-04	11	200	69.61
1.E-04	11.5	200	52.37
1.E-04	12	200	40.34
1.E-04	13	200	25.43
1.E-04	15	200	12.15
1.5.E-04	16.5	200	69.61
1.5.E-04	18	200	40.34
1.5.E-04	19.5	200	25.43
1.5.E-04	21	200	17.12
1.5.E-04	22.5	200	12.15
2.0.E-04	22	200	69.61
2.0.E-04	24	200	40.34
2.0.E-04	26	200	25.43
2.0.E-04	28	200	17.12
2.0.E-04	30	200	12.15

K [m/s]	L [m]	t [h]	H1 [m]
1.E-04	11.5	75	1.05
1.E-04	11.5	100	2.29
1.E-04	11.5	150	10.95
1.E-04	11.5	170	20.47
1.E-04	11.5	185	32.74
1.E-04	11.5	200	52.37
1.E-04	12	75	0.95
1.E-04	12	100	2.01
1.E-04	12	150	9.00
1.E-04	12	170	16.40
1.E-04	12	185	25.72
1.E-04	12	200	40.34
1.E-04	12.5	75	0.87
1.E-04	15.5	100	1.02
1.E-04	12.5	150	7.52
1.E-04	12.5	170	13.38
1.E-04	12.5	185	20.60
1.E-04	12.5	200	31.73
1.E-04	13	75	0.80
1.E-04	13	103	1.73
1.E-04	13	150	6.37
1.E-04	13	170	11.08
1.E-04	13	185	16.79
1.E-04	13	200	25.43

**Table S3.** Left table: possible couples of K and L that produce the same hydraulic head. In green  $\overline{\ }$  are the feasible results,  $\overline{\ }$  and in red  $\overline{\ }$  are the values that are not realistic. Right table: testing different event lengths.