HYD-RESPONSES: daily hydro-meteorological catchment-level time series to analyse HYDrological drought dynamics in RESPONSE to (cumulative) water deficits in Swiss catchments.

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Abstract.

The HYD-RESPONSES dataset ~~(https://doi.org/10.5281/zenodo.14713274; von Matt et al., 2025)~~ (https://doi.org/10.5281/zenodo.14713274; von Matt et al., 2026) provides new daily catchment-level time series for key hydro-meteorological variables necessary to study drought conditions, including precipitation, snow water equivalent, temperature, soil moisture, (potential) evaporation, and streamflow. The dataset covers 184 small to large Swiss catchments of the surface water monitoring network operated by the Federal Office for the Environment (FOEN). The catchments range across a variety of streamflow regime types, mean altitudes, biogeographic regions, and anthropogenic influences. The ~~data set provides daily average streamflow derived from measurements by the FOEN and daily hydrometeorological data (precipitation, temperature, radiation, snow and soil moisture) on the catchment level extracted from spatially gridded data~~ dataset comprises daily mean streamflow observations obtained from the Federal Office for the Environment (FOEN), complemented by daily hydrometeorological variables aggregated at the catchment scale. Complementary variables are derived from from spatially gridded products provided by MeteoSwiss (~~RhiresD, TabsD, TmaxD, TminD, SrelD), MeteoSwiss and~~ Spatial Climate Analyses), the WSL Institute for Snow and Avalanche Research SLF (SPASS), SLF (OSHD), and the European Centre for Medium-Range Weather Forecasts ECMWF (ERA5-Land reanalysis).

In addition, derived indicators describing snowfall, snowmelt, (potential) water balance and streamflow are provided. ~~Information on precipitation, evaporation-driven and streamflow deficits are provided in form of~~ Deficits related to precipitation, evaporation, and streamflow are quantified using both standardized and non-standardized (drought/deficit) indices. Standardized indices include the SPI, SPEI and SMRI and are provided on multiple aggregation scales from 1 to 24 months (mostly in 3-monthly steps). Non-standardized indices are provided as cumulative (water) deficits in (potential) water balance (CWD and PCWD) and streamflow (CQD). For all variables and indices, the climatology and the (standardized) anomalies are available on various time scales (daily, monthly, seasonal, and yearly). Drought event time series containing drought event numbers and drought event durations, are provided for streamflow droughts identified by using two percentile-based event definitions (fixed and variable threshold) and for cumulative water deficits (CWD, PCWD and CQD).

Detailed catchment descriptors covering hydro-climatological and hydro-terrestrial aspects as well as streamflow characteristics are provided for all catchments. The dataset can be used to study weather-driven streamflow extremes, to train

data-driven machine-learning algorithms, to study drought propagation, and for comparative analyses of catchment responses in disturbed and undisturbed catchments. The dataset is compatible with the recently published ~~CAMELS-CH dataset~~ "Catchment Attributes and MEteorology for Large-sample Studies" dataset for hydrological Switzerland (CAMELS-CH) and with additional catchment descriptors provided by the FOEN.

1 Introduction

In recent years, the frequency of droughts has increased in Europe and Switzerland with notable drought years in 2003, 2011, 2015, 2018, 2020. Most recently, in 2022, conditions were characterized as unprecedented in terms of compound heat and drought in the last 500 years over large parts of Europe (BAFU, 2016; BAFU et al. (Eds.), 2019; BUWAL, BWG, MeteoSchweiz, 2004; Scherrer et al., 2022; Tripathy and Mishra, 2023). Under climate change, this trend is likely to continue with projected increases in drought frequency, dry spell duration, and drought severity for both individual and combined drought types (Brunner et al., 2019c, a; Calanca, 2007; Kotlarski et al., 2023; Muelchi et al., 2021a; von Matt et al., 2024). Increasing drought impacts on various sectors are expected. This has prompted Swiss national authorities to establish a national drought early warning system (DEWS, see https://www.trockenheit.admin.ch/en; BAFU (Eds.), 2021; CH2018, 2018; Haile et al., 2020; Henne et al., 2018; Naumann et al., 2021; Brunner et al., 2019a; Otero et al., 2023; Ranasinghe et al., 2021; Tschurr et al., 2020; BAFU, 2022; Swiss Confederation, 2025).

Droughts are an inherently multivariate phenomenon with often non-linear drought propagation from meteorological conditions to impacts on ecosystems, infrastructure, and economy. Individual drought events may differ in their hydro-climatological, hydro-meteorological, hydro-terrestrial and anthropogenic ~~characteristics~~ drivers (Brunner et al., 2023; Hao and Singh, 2015; Mishra and Singh, 2010; Zhou et al., 2021; Floriancic et al., 2020; Massari et al., 2022). The consideration of multiple hydro-climatic, hydro-meteorological, hydro-terrestrial and anthropogenic (disturbance) factors is therefore key to understand catchment-specific drought responses and sensitivities and to provide information for drought early warning, preparations, and interventions (e.g., Apurv et al., 2017; Apurv and Cai, 2020; Baez-Villanueva et al., 2024; Brunner et al., 2022, 2021; Ding et al., 2021; Peña-Angulo et al., 2022; Peña-Gallardo et al., 2019; Sutanto and Van Lanen, 2022; Tijdeman et al., 2018; Van Lanen et al., 2013; Savelli et al., 2022; Van Loon and Laaha, 2015; von Matt et al., 2024).

Novel high-resolution observational datasets provide a unique opportunity to combine multiple hydro-meteorological variables to analyze and monitor drought dynamics and the evolution of drought impacts of individual events at the catchment-level. For example, the propagation of meteorological to hydrological droughts or the evolution of droughts from the development to the recovery phase can be studied (Brunner et al., 2021; Brunner and Chartier-Rescan, 2024; Parry et al., 2016; Raposo et al., 2023; Brocca et al., 2024a; Brunner et al., 2021; Stocker et al., 2023; Poussin et al., 2021). The Federal Office for Climatology and Meteorology (MeteoSwiss) provides a suite of high-resolution essential climate variables spatially interpolated to a regular grid from a dense measurement station network (MeteoSwiss, 2024). Further, new high-resolution snow climatologies produced by both MeteoSwiss and the WSL Institute for Snow and Avalanche research SLF have recently become available, providing a

novel opportunity to analyze the long-term influence of snow processes, which are crucial for streamflow (drought) generation in Alpine catchments in Switzerland (Staudinger et al., 2014, 2017; Avanzi et al., 2024; Brunner et al., 2023; Koehler et al., 2022; Michel et al., 2023; Marty et al., 2025).

Observation-based evapotranspiration and soil moisture data is sparse in Switzerland. Hence, information on these variables is often extracted from hydrological model simulations Brunner et al. (2021); Melsen and Guse (2019); Samaniego et al. (2013, 2018). The ERA5-Land reanalysis dataset, provided by the European Centre for Medium-Range Weather Forecasts (ECMWF) (Muñoz-Sabater et al., 2021), offers a compromise between high spatial resolution and long temporal coverage and is better suited for hydro-meteorological analyses and modelling over more complex terrain such as Switzerland than the ERA5 reanalysis datasets (Muñoz-Sabater et al., 2021). A frequently used approach for analyzing drought propagation from meteorological (precipitation) to agricultural (soil moisture) and hydrological (streamflow and/or groundwater) droughts relies on standardized drought indices based on e.g., precipitation and/or evaporation (by using the standardized precipitation index (SPI) or the standardized precipitation evaporation index (SPEI) (Raposo et al., 2023; Barker et al., 2016; Peña-Gallardo et al., 2019; Zhou et al., 2021). These standardized drought indices are typically aggregated over varying retrospective time scales (months to years) and are useful proxies for various factors that determine catchment-scale water balances, including soil moisture, streamflow, groundwater, and snow processes (Bachmair et al., 2018; Tschurr et al., 2020; European Commission, 2020; Cammalleri et al., 2019; Staudinger et al., 2014). Longer aggregation scales hereby reflect response scales of storage components with longer memory, while shorter scales reflect streamflow and/or soil moisture in smaller catchments, mainly influenced by pluvial processes (Bachmair et al., 2018; Baez-Villanueva et al., 2024; Haslinger et al., 2014; Myronidis et al., 2018; Staudinger et al., 2014; Tschurr et al., 2020; WMO and GWP, 2016; Yihdego et al., 2019; Cammalleri et al., 2019; Bachmair et al., 2016; European Commission, 2020). Standardized drought indices are now widely used in DEWS (Bachmair et al., 2016; Kchouk et al., 2022; Raposo et al., 2023; Tijdeman et al., 2020) and will also be used in the Swiss DEWS (L. Benelli, pers. comm.).

Recent studies focused on assessing the benefits of non-standardized (deficit) indices in tracking the drought propagation signal across drought types (see e.g., Brunner and Chartier-Rescan, 2024; Sur et al., 2020; Wu et al., 2020). Non-standardized indices provide physically interpretable and consistent information on deficits which remain inter-comparable across systems ~~as a result of non-transformation~~ (Van Loon, 2015; Raposo et al., 2023; Wu et al., 2020). Examples are the Hydrological Anomaly Index (HAI), the Water Balance Drought Index (WBDI), the cumulative water deficits (CWD), and the potential cumulative water deficit (PCWD) (Stocker et al., 2023; Sur et al., 2020; Wu et al., 2020). Non-standardized indices allow direct quantification of (precipitation) deficits or surpluses associated with the drought propagation into and recovery from a (hydrological) droughts (Wu et al., 2020) and hence provide valuable information for proactive water management and decision-making (Xu et al., 2023; Parry et al., 2018).

Here, we present a novel dataset with high-resolution observational daily catchment-level time series for key hydro-meteorological variables (including precipitation, snow water equivalent, temperature, soil moisture, (potential) evaporation and streamflow), standardized and non-standardized (drought/deficit) indices (SPI, SPEI, SMRI, CWD, PCWD, CQD) and (streamflow) drought events covering 184 small to large catchments in Switzerland. The HYD-RESPONSES

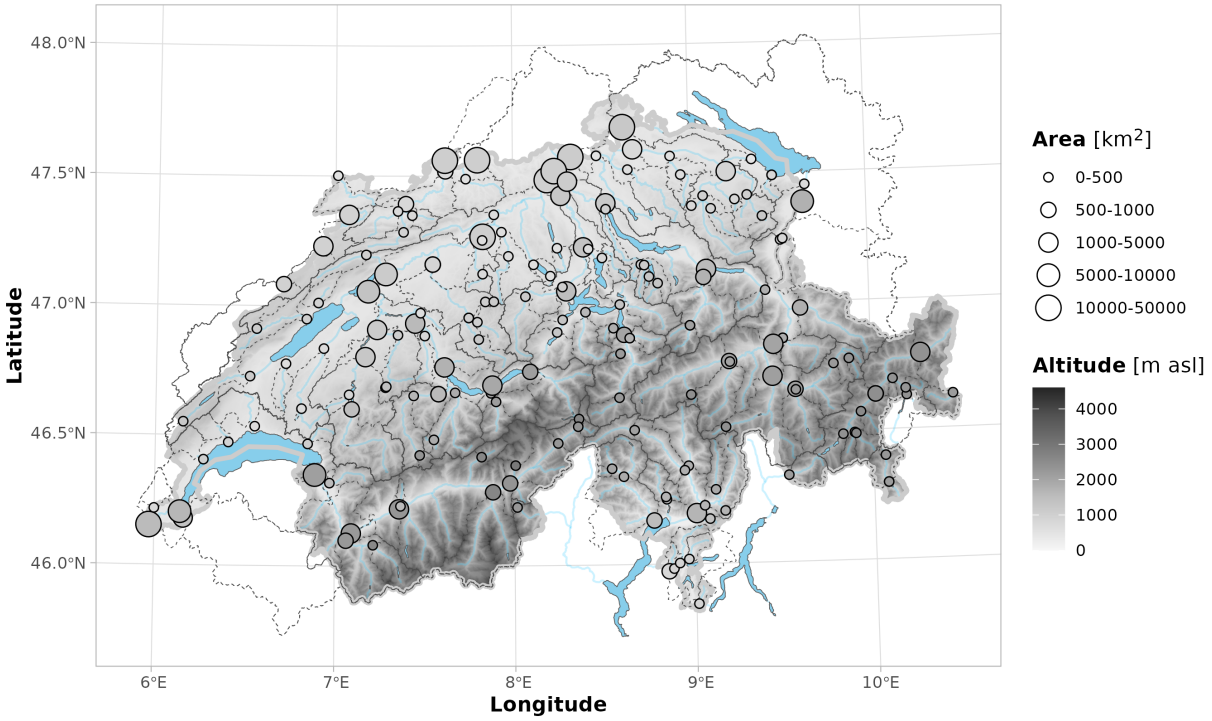

Figure 1. Overview of the study area and catchments included in the HYD-RESPONSES dataset. Catchment outlets (circles) are coloured by mean catchment altitude [m a.s.l.] and the point size scales with the catchment area [km^2]. Dashed lines show the catchment outlines. Generalized streamflow networks and lakes are shown in light blue.

dataset can be combined with existing hydro-meteorological time series datasets and catchment descriptors such as CAMELS-CH (Höge et al., 2023a)—, which provides large-sample hydro-meteorological data for hydrologic Switzerland and is the Swiss version of the "Catchment Attributes and MEteorology for Large-sample Studies" (CAMELS; see e.g., Clerc-Schwarzenbach et al., 2024). The remaining paper is structured as follows: In section 2 the study region and the catchments are presented. Section 3 introduces all datasets used to compile the HYD-RESPONSES dataset. Section 4 elaborates on the processing of hydro-meteorological data. Section 5 is the analogue for the processing and extraction of catchment descriptors. Section 6 finally discusses the dataset and points to potential caveats and cautionary notes while section 7 presents multiple complementary datasets which are valuable in combination with the HYD-RESPONSES dataset. Section 8 provides a concluding summary. Three exemplary use cases to illustrate the nature and potential of the HYD-RESPONSES dataset are provided in Appendix A.

2 Study region and catchments

The 184 catchments (Fig. 1) provided in the HYD-RESPONSES dataset span a wide range of catchment areas (0.56–35'878 km^2), glaciation percentages (0–56 %), altitude ranges (467–2937 m a.s.l.) and 18 streamflow regime types (~~n=18) (~~see Fig. 2). ~~More than half~~ A bit more than half of the catchments 51 % (n=94~~(51 %)) of the catchments)~~ are small to ~~mid-sized~~ mid-size with an area of between 10 km^2 and 500 km^2. Only 9 (4 %) catchments are smaller than 10 km^2 and 56 (30.4 %) catchments are larger than 500 km^2. The dataset contains eight very large catchments with areas between 10000 km^2 and 50000 km^2 (max. area = 35'878 km^2), associated with the three largest rivers in Switzerland: Aare, Rhine and Rhone. Most catchments (82.5 %) have less than 5 % glaciated area. ~~The~~ In terms of mean catchment height (elevation), the catchments are distributed relatively equally between 500 and 2500 m a.s.l. ~~with fewer (77 out of 98)catchments at elevation ranges above 1500 m a. s.l..~~ (see Fig. 2c). Only eight catchments are higher than 2500 m a.s.l. and only one catchment is at very low elevation (catchment Wiese, Basel). Streamflow regime types were classified and adjusted by the FOEN based on data from the Hydrological Atlas of Switzerland Table 5.2 (https://hydrologischeratlas.ch/downloads/01/content/Tafel_52.pdf). Catchments smaller than 500 km^2 are characterized by considering mean altitude and catchment glaciation percentage to reflect the contribution of specific streamflow (drought) generating processes (glacial, nival, pluvial). Catchments larger than 500 km^2 are generally classified as *mixed regime (>500 km^2)* type and contain catchments characterized by a combination of streamflow (drought) generating processes. For more information see also Aschwanden and Weingartner (1985) and Fig. 2e.

Note that 12.5 % (n=23) of the catchments have at least 5 % of catchment area lying outside of the Swiss national borders as the dataset consists of catchments of the entire hydrological Switzerland (catchments that drain in(to) Switzerland). Furthermore, the Swiss streamflow monitoring network is designed such that multiple measurement stations may be located along the same river. As a result, upstream catchments can be nested within larger downstream catchments, leading to hierarchical dependencies.

3 Input data products

In this section, the input datasets used to produce and compile the HYD-RESPONSES dataset are presented and reference literature for further reading and more detailed information is provided. Original data products are provided by the Federal Office for Climatology and Meteorology (MeteoSwiss), the Federal Office for the Environment (FOEN), the Swiss Federal Office of Topography (Swisstopo), the Federal Office for Agriculture (FOAG), the WSL Institute for Snow and Avalanche Research (SLF) and the European Centre for Medium-Range Weather Forecasts (ECMWF).

3.1 Catchment-level time series data from streamflow observations

Daily average streamflow measurements at the catchment outlet were provided by the FOEN via the Hydrological Service (www.hydrodaten.admin.ch) for more than 200 stations. The data availability is station-specific and depends on the installation and FOEN-internal data quality checking. The HYD-RESPONSES dataset only provides a subset of 184 catchments

Table 1. (Spatially gridded) Data products used for the time series extraction. Full variable names and associated units are provided in Table B1 (glossary) in Appendix B. Note that the variable short names correspond to the layer/product names in the respective dataset.

Dataset	Variables	Period	Spatial resolution	Temporal resolution	Producer
Spatial Climate Analyses	TabsD, RhiresD TminD, TmaxD, SrelD	1961–2023 1971–2023	1×1 km	daily	MeteoSwiss
Snow Climatology for Switzerland (SPASS)	SWECLQMD	1961–2022	1×1 km	daily	MeteoSwiss & SLF
Climatological snow data since 1998 (OSHD)	swee, romc	1998–2023	1×1 km	daily	SLF
ERA5-Land	tp, t2m, e, pev, smlt, sd, ssr, ro, sro, swvl1, swvl2, swvl3, swvl4	1950–2023	0.1×0.1°(ca. 9×9 km)	hourly	ECMWF
Streamflow time series	Q	Station specific	catchment-level (outflow point data)	daily	FOEN

by considering and considers only stations for which an analysis of hydrological drought dynamics in response to cumulative water deficits was deemed to be meaningful in correspondence with the FOEN (Caroline Kan; see Fig. 1). Stations were excluded in case of i) Q measured at Namely stations which provide reliable streamflow (Q) time series and are associated with a physical/natural catchment. Stations have therefore be excluded if they i) only provide water-level stations (information (no Q, 3 stations), ii) Q measured at NADUF-stations (are not part of the main streamflow measurement network (e.g., stations from other networks such as the National Surface Water Monitoring Programme (NAWA BAFU, 2023), 4 stations), iii) secondary stations (11 stations), iv) stations with potential return streamflow (= negative) streamflow (Q values, 2 stations), v) Q measured at derivations (2 stations), vi) stations with no without watershed delineation (i.e., subterranean; 1 station) and vii) uncertainties in time series composition due to displacement and/or temporarily missing Q of contributing stations (4 stations). A The complete list of included stations is provided in Tables ??, ??, ?? and ?? (Appendix)B6, B7, B8, B9 and B10 in Appendix B.

3.2 Catchment-level time series data derived from spatially gridded products

Meteorological variables (except for evaporation)were assembled Hydro-meteorological variables used in this study were compiled from multiple complementary data sources, combining station-based spatial climate analyses, dedicated snow model products, and reanalysis data (Table 1). This multi-source approach allows both comprehensive coverage of relevant variables

and comparative analyses between different data products.

Meteorological variables derived from station-based observations were obtained from the high-resolution (1×1 km) spatial climate analyses provided by MeteoSwiss (MeteoSwiss, 2024)(see Table 1). The variables . These include average 2 m temperature (TabsD), daily minimum and maximum 2 m temperature (TminD, TmaxD), daily precipitation sums (RhiresD), and daily sunshine duration (SrelD) (Frei, 2014; Frei and Schär, 1998; MeteoSwiss, 2021a, b, c). The data Data availability is product-specific and covers the period 1961–2023 1961—2023 for RhiresD and TabsD and 1971–2023 for the other products (1971—2023 for TminD, TmaxD, SrelD)and SrelD. The spatial climate analyses products used here only generally cover the Swiss territory , except for only, with the exception of RhiresD, which covers catchments located outside Switzerland , but draining also includes catchments outside Switzerland that drain through Swiss territory. Note that RhiresD is not available for catchments covering regions in France and Italy before 1992 due to limited meteorological station availability and hence limited reduced data reliability (MeteoSwiss, 2021a). Catchments with a significant area in France or Italy may therefore be handled with care and/or potentially be excluded from analysis before substantial areas in these regions should therefore be treated with caution or excluded from analyses prior to 1992 (see Section ??Sect. 6.2).

Snow water equivalent (SWE) data was were compiled from two independent high-resolution (1×1 km) snow datasets. The first and main product resulted from the joint research project "A primary source is the spatial Snow Climatology for Switzerland (SPASS) " developed jointly by MeteoSwiss and SLF (Michel et al., 2023; Marty et al., 2025). The preliminary version was produced in 2022 and This dataset provides modelled and bias-corrected daily SWE data for the period for September 1961–September 2022. The spatial extent is restricted to the Swiss territory. The SPASS SWE is based on the daily 2022, derived using the SnowQM model based on TabsD and RhiresDproducts (see above) and makes use of a quantile-mapping approach. The . The SnowQM model is presented in detail in Michel et al. (2023). The second snow product is based on spatial coverage is restricted to Switzerland. A second snow dataset is derived from the Swiss Operational Snow-hydrological Snow-Hydrological model system (OSHD)and is provided by the , provided by WSL (SLF)(Mott, 2023; Mott et al., 2023). The OSHD data provides information on both SWE and , which supplies SWE and modelled snowmelt runoff for the period 1998– 2022 1998–2022 (Mott, 2023; Mott et al., 2023).

All other remaining hydro-meteorological variables, including evaporation, potential evaporation, soil moisture, and additional variables already covered by the previously introduced datasets overlapping with the datasets described above, were extracted from the ERA5-Land reanalysis dataset provided by ECMWF (Muñoz-Sabater et al., 2021). Several variables are therefore covered by multiple source data and are all included thus available from multiple sources and are retained in the HYD-RESPONSES dataset to allow comparative analyses between the different data products. Time series covered by multiple data sources include temperature variables enable inter-product comparisons. Variables covered by more than one data source include air temperature (TabsD, TminD and TmaxD from MeteoSwiss, t2m from ERA5-Land), precipitation (RhiresD from MeteoSwiss, precipitation from ERA5-Land), potential and total evaporation (ERA5-Land), sunshine duration (SrelD),

snow water equivalent (SWE; from SPASS, OSHD, and ERA5-Land) , modelled snow melt and modelled snowmelt (from OSHD and ERA5-Land)and streamflow (FOEN). Additional variables extracted from ERA5-Land include four . Additional ERA5-Land-specific variables include soil water volume levels (swvl) at four depths, total solar radiation (ssr)and , total runoff (ro), and surface runoff (sro). For a more detailed description of variables, see the data Detailed variable descriptions are provided in the dataset documentation on Zenodo (von Matt et al., 2025)(von Matt et al., 2026). A glossary of for the variable abbreviations is provided in Table B1.

The ERA5-Land data is provided at an hourly temporal are available at hourly resolution for the period 1950–2023 and can be accessed 1950—2023 via the Copernicus climate data store Climate Data Store (CDS) (https://cds.climate.copernicus.eu/datasets/reanalysis-era5-land). We preferably included data from ERA5-Land over data from consists of numerical model output from the ECMWF land surface model which itself is driven by downscaled and elevation-corrected meteorological forcing from ERA5 due to the (Muñoz-Sabater et al., 2021). The higher spatial resolution of ERA5-Land (0.1×0.1°, ca. approximately 9×9 km) . To ensure consistency with the other hydro-meteorological input datasets, the hourly results in an enhanced soil moisture representation and river discharge estimations making ERA5-Land data was aggregated to daily values (see Section 4.1). more suitable for analyses based on the hydrological cycle than ERA5 (Muñoz-Sabater et al., 2021).

Overview of the spatial raster products used to extract daily time series. **(a)** Mean daily temperature (TabsD, MeteoSwiss), **(b)** Daily precipitation sun (RhiresD, MeteoSwiss), **(c)** Daily snow water equivalent of the Swiss snow climatology (SPASS) (SWE, MeteoSwiss & SLF), **(d)** Daily evaporation sum (aggregated from hourly ERA5-Land data, ECMWF). Note that the second snow climatology product (OSHD) is not shown. Contours in white/black show catchment *2034 - Broye, Payerne, Casernde d'aviation* for the day with the highest observed catchment average values for each specific product for the year 2022. White squares show the catchment outlet where daily streamflow is measured. Extracted and derived time series over the year 2022 are shown for the same catchment in Figure 6. The procedures used to extract time series from all gridded datasets are detailed in Section 4.

3.3 Catchment-level time-invariant data (catchment descriptors)

Datasets used to compile an extensive set of catchment descriptors include station metadata and information on time series availability and homogeneity provided by the FOEN as well as spatial (polygon) data on hydro-terrestrial characteristics (e.g., soil characteristics, hydro-geology) provided by the FOEN, FOAG and Swisstopo (see Tab. Table 2). Most information is available from www.opendata.swiss, the FOEN Hydro-Service (www.hydrodaten.admin.ch, or can be downloaded and inspected via www.map.geo.admin.ch (Swisstopo). Direct links to the datasets are provided below in section 10.and in section *Code and data availabilty*.

Table 2. Data products used to extract catchment descriptors.

Dataset	(Extracted) Variables	Producer
Digital soil suitability maps of Switzerland	soil wetness, soil depth, permeability, water holding capacity, nutrient content and skeletal content	FOAG
Hydrogeological map of Switzerland	aquifer type (loose or solid rock), aquifer genesis and aquifer productivity	FOEN
Lithological map for Switzerland	dominant rock type classes (loose, sedimentary and crystalline rock)	Swisstopo
Springs and swallow holes in karst regions	number of springs (per km^2)	FOEN
~~swissALTI3D (DEM)aspect, slope Swisstopo~~swissTLM3D Hydrography	Drainage density	Swisstopo
Biogeographic regions of Switzerland	Biogeographic regions	FOEN
Catchment metadata	time series availability, breakpoint analysis, area, mean height, outlet coordinates and streamflow regime type	FOEN

The digital soil suitability maps of Switzerland provide information on a set of different soil characteristics assessed on 25 different geological and geomorphological units which are further discriminated by different landscape elements depending on aspect, slope and bedrock. The maps were first assessed in 1980 and revised in 2000 (BLW, 2022; Swisstopo, 2020). The different soil characteristics include soil wetness, soil depth, permeability, water storage capacity, nutrient content and skeletal content.

The hydro-geological map of Switzerland provides information on groundwater resources in Switzerland (Schürch et al., 2007), including information on aquifer type (loose or solid rock), aquifer genesis and aquifer productivity. The map was originally produced and published for the Hydrological Atlas of Switzerland (HADES, https://hydrologischeratlas.ch/). The hydro-geological information was further complemented with the lithological map for Switzerland (produced by Swisstopo), which provides a general overview of dominant rock type classes (loose, sedimentary and crystalline rock). The maps are available via opendata.swiss (hydrogeological map, lithological map) or can also be accessed via the Hydrological Service of the FOEN (https://www.bafu.admin.ch/bafu/de/home/themen/wasser/zustand/karten/geodaten.html). The number of springs and swallow holes in karstic regions provides additional information related to aquifers and the contribution of subsurface water storage. The layer provides main discharge source locations in karstic regions and is available via opendata.swiss (produced by FOEN). ~~Standard topographical characteristics such as slope and aspect were derived from the high-resolution digital elevation model (swissALTI3D) publicly available via Swisstopo at a resolution of 2 m (Swisstopo, 2022).~~ The swissTLM3D Hydrography provides topological information on the different water bodies of Switzerland (including flowing and stagnant waters) and originates from the swissTLM3D dataset provided by and accessible via Swisstopo.

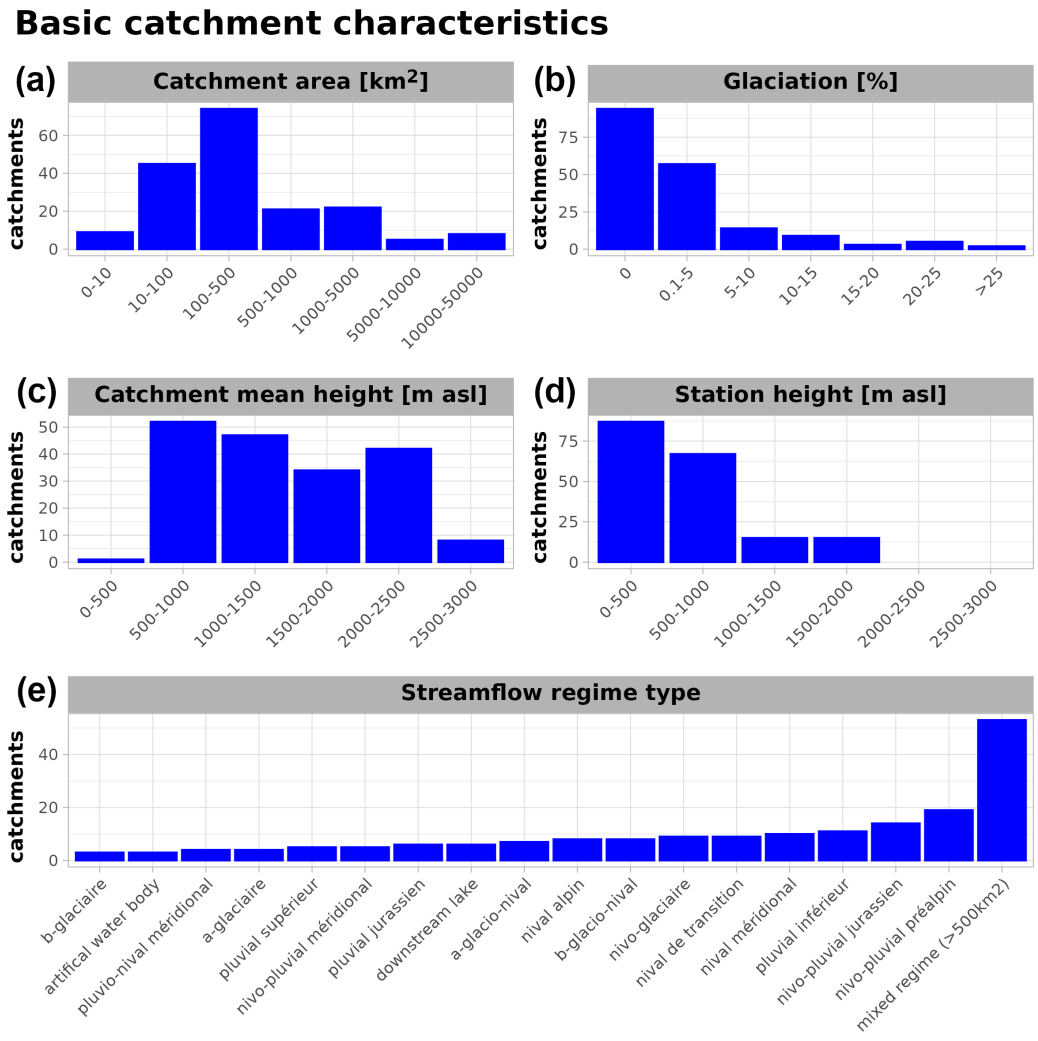

Figure 2. General catchment characteristics provided by the FOEN. **a)** Catchment area in km^2, **b)** Glaciation percentage (of catchment area), **c)** catchment mean height [m a.s.l.], **d)** height of the streamflow gauge measurement station [m a.s.l.] and **e)** streamflow regime types. The Y-axis shows the ~~frequency~~ number of catchments in each category.

The biogeographic regions of Switzerland provide six regions differentiated by similarity of flora, fauna, bryophytes and ornithological information as well as homogeneous surface water catchments (BAFU (Eds.), 2022). Biogeographic (eco-)regions often correspond well to catchment groups with similar streamflow regime types and are therefore frequently used for catchment regionalization (e.g., Jehn et al., 2020; Guo et al., 2021). The biogeographic regions are available via opendata.swiss.

Finally, general information on the gauging stations and streamflow time series (availability and homogeneity) were provided as accompanying (meta-)data by the FOEN. Time series homogeneity was ~~assessed by a FOEN-internal breakpoint analysis for time series homogenization~~ derived by the FOEN using breakpoint tests following the method of Bai and Perron (1998) . Breakpoints are identified by partitioning the time series based on the number of potential breakpoints and subsequent modeling of the time series by piecewise linear regression (see Bai and Perron, 1998). The optimal breakpoints are found by minimization of the sum of squared residuals. Resulting breakpoints are indicative to changes in the mean annual 7d mean flow (M7Q) and were further plausibilized by the FOEN based on catchment meta information and known (potentially) relevant anthropogenic influences such as the construction of (reservoir) dams, hydropower and wastewater treatment plants (for more information see BAFU, 2024). General station information includes catchment area, mean ~~height~~catchment height (elevation), glaciation percentage, outlet coordinates and streamflow regime type (among others) (see Figs. 1 and 2). Catchment outlines (polygons provided by the FOEN) and catchment outlets (point shapes) are provided in the coordinate system CH1903/LV03 (EPSG:21781).

Note that the digital soil suitability maps, swissTLM3D hydrography, biogeographic regions of Switzerland as well as information on springs and swallow holes in karst regions are restricted to Swiss national territory. Catchments with a significant area outside of Switzerland should be treated with caution regarding descriptive variables extracted from these datasets (see Sect. 5 for a comprehensive overview on extracted descriptors). The hydrogeological and lithological maps of Switzerland to a large extent also cover areas outside of Switzerland. Only catchments of the Rhine (catchments 2091, 2143, 2288, 2289) and Wiese (catchment 2199) are not entirely covered. However, with a coverage of at least >94%, descriptors extracted from these datasets may still prove valuable for these catchments.

Methodological details on the extraction and preparation of catchment descriptors are presented in Section 5.

4 ~~Data processing~~Processing of hydro-meteorological data

This section describes the methodology used for aggregating spatially gridded ~~data products and catchment descriptors on the catchment level~~(time series) data product, the methods used to derive additional indicators, standardized drought indices, and presents the definition and declaration of (hydrological) drought events. Guidance on the reliability of indicators is provided through a three-level classification based on the origin of the underlying data, the extent to which variables rely on (model) assumptions, and the degree of processing applied to derive the hydro-meteorological data, (drought) indices, and events. *Level 1* consists of direct, unaltered measurement data and is therefore considered the most reliable. *Level 2* includes data directly extracted from publicly available spatially interpolated hydro-meterological datasets and data subjected to only minimal (post-)processing (e.g., temporal aggregation). *Level 3* comprises all variables, indicators, and indices derived by the authors, irrespective of the degree of validation or verification performed. For both *Level 2* and *Level 3* data, additional annotations are provided for variables whose derivation is based on a (strong) modeling component. A summary of the classification is

Table 3. Three-level reliability classification used for hydro-meteorological data in the HYD-RESPONSES dataset.

Level	Description
Level 1 (L1)	Direct, unaltered measurement data. These data are considered the most reliable, as they are not subject to interpolation, modeling assumptions, or additional processing.
Level 2 (L2)	Data extracted from publicly available, spatially interpolated hydro-meteorological datasets and data subjected to only minimal post-processing, such as temporal aggregation. Variables whose derivation relies on strong underlying (modeling) assumptions are explicitly annotated.
Level 3 (L3)	Variables, indicators, and indices derived by the authors, irrespective of the degree of validation or verification performed. Variables whose derivation relies on strong underlying (modeling) assumptions are explicitly annotated.

provided in Table 3.

The naming of the unaltered variables directly retrieved from measurement data (streamflow) or extracted from spatially interpolated hydro-meteorological datasets is based on the layer names used in the original input datasets (see variables listed in Table 1 and the glossary provided in Table B1). Derived variables and (standardized) drought indices are named by a suffix representing the type of indicator followed by all contributing variables where ERA5-Land variables are kept in lowercase while variables from other products start with upper-case letters. Naming for derived variables based on the snow products make use of the product name (SPASS) or the combination of product and variable names (OSHD) as identifier for clear distinction. All extracted and derived variables and their suggested reliability level are listed in the Tables B2, B3, B4 and B5 in Appendix B.

Time series of all categories are illustrated in Fig. 6 showing the exceptional drought year 2022 for the example catchment *2034 - Broye, (Payerne, Caserne d'aviation)* located in the western Swiss Plateau region (see catchment contours in Fig. A4). A detailed analysis of the event year 2022 demonstrating the utility of the time series provided in the HYD-RESPONSES dataset is provided in Sect. A2 in Appendix A.

4.1 Time series extraction

Based on the spatially gridded hydro-meteorological input products (see ~~Section~~ Sect. 3.2), catchment-level time series were extracted using the R-packages *terra* (Hijmans, 2023) and *exactextractr* (Baston, 2023). First, the hourly ERA5-Land data was aggregated to daily resolution following the standards used by the MeteoSwiss spatial climate analyses (e.g., RhiresD and TabsD). For this, instantaneous ~~and accumulation/flux variables~~ variables and variables representing accumulations or fluxes are distinguished. For instantaneous variables, we provide daily average values. For ~~accumulation and flux~~variables

representing accumulations and fluxes, we provide ~~variables~~ daily sums. Flux variables (mainly precipitation and evapotranspiration) ~~were further aggregated consistently with~~ are aggregated using the same temporal convention as the RhiresD precipitation sums, i.e., from 06 UTC (day) to 06 UTC (day + 1) (see MeteoSwiss, 2021a). Instantaneous variables and ERA5-Land temperature were averaged from 00 UTC to 00 UTC ~~, which is consistent with the other~~ again following the convention used in equivalent MeteoSwiss products (e.g., TabsD; MeteoSwiss, 2021b). Daily catchment-average time series were then extracted by using the catchment outlines (polygons) provided by the FOEN. Units were homogenized across time series. ~~The units~~ Both units and full standard variable names are listed in Table B1 (glossary).

The length of the time series depends on the dataset that they were derived from (see Table 1 for details). Streamflow time series are provided for three different catchment-specific time periods: (1) the original time series (entire period), (2) the most recent gap-free time-period time series and (3) the most recent homogeneous time series (in case of significant and plausible breakpoints; otherwise equal to the gap-free time series) (see Fig. 3). The breakpoint information is provided by FOEN ~~(for more information see BAFU, 2024).~~ (see Sect. 3.3). Information on the start of the streamflow monitoring by ~~limnographs~~ water level sensor is also provided. The streamflow data should only be considered reliable after the initialization of a ~~limnograph~~sensor. In case of no breakpoints the gap-free period is equal to the homogeneous period. The homogeneous period is usually the shortest (e.g., in case of breakpoints or ~~limnograph~~ water level sensor initialization; see for example catchment 2349 in Fig. 3). In the case of gaps but no breakpoints, both the homogeneous and the gap-free periods are identical (see, i.e., catchments 2239, 2386 and 2368 in Fig. 3). Indicators and (non-)standardized (drought/deficit) indices derived from the hydro-meteorological time series are available for the longest common period of all contributing variables.

4.2 Derived indicators

4.2.1 Streamflow

Derived indicators related to streamflow consist of the 7-day average streamflow (moving average) M7Q ~~.~~ (Fig. 6i). The M7Q (or M7) is often used in low-flow studies and is also used for the official low-flow statistics in Switzerland by the FOEN (see e.g., BAFU, 2024; Muelchi et al., 2021a; von Matt et al., 2024).

4.2.2 Snow related variables

In addition to variables providing direct information on (modelled) snowmelt, also daily differentiated SWE (ΔSWE) time series are provided for both SPASS and OSHD. Snowfall ($\Delta SWE > 0$) and snowmelt ($\Delta SWE < 0$) time series are provided separately. Note that ~~the SPASS~~ SWE is reset in the SPASS dataset at the end of every snow year (every September 1st~~st~~) to avoid unrealistically high ~~snow water equivalent accumulation ("snow towers")(Michel et al., 2023). This~~ accumulation of snow water equivalents ("snow towers"; see Michel et al., 2023). As snowfall and snowmelt were derived from daily differences in SWE (ΔSWE), this reset can result in ~~large snowmelt amounts (~~an artificially large negative ΔSWE ~~< 0) around~~ value on September 1st~~.~~ ~~st~~ that does not represent actual physical snowmelt. To prevent this model artifact from affecting the derived

snowmelt time series, ΔSWE values on September 1~~st were therefore~~st were set to missing values and replaced by a linear interpolation ~~between the day before and the day after~~using the ΔSWE values from the preceding and following days. Snow-corrected precipitation series (P+ ΔSWE) were calculated by combining time series of total precipitation (RhiresD and ERA5-Land) and ΔSWE time series (SPASS, OSHD) as well as time series with modelled snowmelt information (SPASS, OSHD and ERA5-Land). Negative snow-corrected precipitation amounts (e.g., RhiresD < ΔSWE) were set to zero.

4.2.3 Water balance

(Potential) Water balance indicators (P–E and P–PET) were derived by combining the total and snow-corrected precipitation time series with the ERA5-Land evaporation and potential evaporation time series.

4.3 Cumulative water deficits

~~Cumulative (potential)~~ (Potential) Cumulative water deficits (CWD and PCWD) are non-standardized indicators tracking evaporation-driven deficits in the (potential) water balance. CWD and PCWD were derived from the daily water balance indicator time series (see ~~Section ??~~Sect. 4.2.3) using the *cwd* R-package (Stocker et al., 2023; Stocker, 2021). A deficit starts when the water balance is negative (i.e., ~~$P - E < 0$~~$P - E < 0$) and is accumulated as long as the deficit remains uncompensated (deficit > 0). Note that no surplus information is tracked. Once the deficit is compensated, the values remain at zero (CWD $= 0$). Example time series are shown for both CWD and PCWD in Fig 6k,l. In some cases ~~, PCWDs~~ (especially for P–PET based only on ERA5-Land variables~~)~~, i.e., tp−pev), PCWDs are not compensated each year and can persist over multiple years. Both CWDs and PCWDs are hence also provided on a yearly calculation basis (annual reset on December 31~~st~~st). Non-standardized indices preserve units (here millimetres) and are physically interpretable in terms of absolute deficit amounts. Cumulative water deficits do not rely on a predetermined calculation time window, which allows the user to track both deficits accumulated over short periods in time (below one month) and deficits accumulated over very long periods.

Water-balance based non-standardized drought indices are widely in use in ecohydrological, land-atmosphere interaction research and catchment-memory studies both with and without temporal resets (see e.g., Biegel et al., 2025; Cui et al., 2022; Stocker et al., 2023). Being more strongly tied to the actual physical water availability, non-compensated CWDs may provide valuable information on carry-over effects in multi-year drought contexts and/or long-term shifts in climatic water balance (Stocker, 2021; Fowler et al., 2022; Saft et al., 2015). PCWDs in contrast are based on potential water balance and (absolute) carry-over deficits should hence be treated with caution. CWD and PCWD time series which are annually reset provide complementary year-to-year information, which may better align in contexts of annual low-flow statistics and allows for a year-to-year comparison across years and catchments.

4.4 Standardized (drought) indices

Standardized (drought) indices depict the anomaly of a deficit over a fixed retrospective ~~period~~ time window (e.g., 1 month). The ~~hydrometeorological~~ hydro-meteorological indicator time series is first aggregated over the given ~~period~~ window and

then transformed to a standard normal distribution by fitting a suitable candidate distribution (Tijdeman et al., 2020; Stagge et al., 2015). Standardized indices therefore provide information on both anomalously dry and wet conditions, which are often defined by thresholds corresponding to standard deviations (STD). As such, values below -1 STD indicate drier than normal conditions (moderate droughts), while values above $+1$ STD indicate wetter than normal conditions (moderate wetness) (McKee et al., 1993; Tschurr et al., 2020). The HYD-RESPONSES dataset provides daily time series for three standardized (drought) indices: the Standardized Precipitation Index (SPI, McKee et al., 1993), the Snowmelt and Rain Index (SMRI, Staudinger et al., 2014), and the Standardized Precipitation Evaporation Index (SPEI, Vicente-Serrano et al., 2010). ~~SPI and SMRI represent precipitation-driven deficits , as they are based on total (SPI; Ponly)or snow-corrected (SMRI ;~~ The SPI represents deficits driven by precipitation only (derived from P), while the SMRI tracks deficits in liquid water input originating from both rainfall and snowmelt (derived from P+ ΔSWE) ~~precipitation time series~~accounting for seasonal snowfall and snowmelt dynamics (Staudinger et al., 2014; Baez-Villanueva et al., 2024). The SPEI ~~accounts for~~ represents deficits driven by ~~evaporation and is derived from the potential water balance (P–PET)~~ evaporative demand (derived from P-PET) and hence indirectly accounts for temperature effects (Vicente-Serrano et al., 2010; Mwinjuma et al., 2026; Gebrechorkos et al., 2025) . Daily time series for all three indices (SPI, SPEI, SMRI) are provided for aggregation ~~periods~~ windows ranging from 1–24 months (31–730 days). Exemplary SPI and SMRI time series for all aggregation windows are shown in Fig. 6g,h.

All indices were calculated using the *SCI* ~~-package~~ R-package (Stagge et al., 2015; Gudmundsson and Stagge, 2016) with custom modifications accounting for the daily time series resolution. All candidate distributions provided within the *SCI* ~~-package~~ R-package (*gamma, genlog, gumbel, lnorm, norm, gev, pe3, weibull*) were tested for suitability. The distributions were fitted for each day of the year (DOY) based on the reference period 1991–2020. This results in a fit for each DOY derived from the same (window of) values for each distribution. Monthly SPI fits (SPI-1) are for example based on the 30 daily values up to the specific DOY for each of the 30 years in the reference period 1991–2020. The suitability of candidate distributions was assessed based on three indicators: the Shapiro-Wilks normality tests (p-values; Shapiro and Wilk, 1965), the number of flags returned by the fitting function ~~(usually indicating convergence issues)~~*fitSCI* (see *SCI* R-package; Gudmundsson and Stagge, 2016), and the number of missing and/or implausible values. Implausible values are defined as values above or below ~~+3 (−3)~~ ± 3 STD following Stagge et al. (2015). Estimating more extreme standardized index values from a 30-year climatology requires substantial extrapolation of the fitted distribution and is therefore associated with large uncertainty, particularly given the strong temporal autocorrelation of drought indices. Values beyond ± 3 correspond to events with return periods far exceeding the length of the reference record and cannot be robustly quantified (see Stagge et al., 2015).

The returned flags in distribution parameter fitting were mainly related to convergence issues (non-convergence) (flag 3, see SCI R-package Gudmundsson and Stagge, 2016). Without a valid fit, the transformation to standardized index values is not possible resulting in missing values on the flagged DOYs in all time series years. As in Staudinger et al. (2014), one best-fitting distribution (over all DOYs) is chosen for all catchments ~~and~~ to allow for catchment comparability. The distribution was selected among the distributions satisfying the following conditions: (1) the transformed values are not significantly different from a normal distribution for the majority of catchments (p-values> 0.05 for at least 75 % of the catchments), (2) fewer than 5 DOYs flagged and (3) fewer than 50 implausible and/or missing values in the transformed

time series (combined consideration of missing values due to flags and unrealistically high/low values). The distribution selection procedure is illustrated for the SPEI in Fig. 4. The results of the Shapiro-Wilks tests (p-values) and information on missing/implausible values and flags are also provided in the HYD-RESPONSES dataset and can be used to identify catchments with non-satisfying properties within the overall best-fitting distribution (see Fig. 4). Following Stagge et al. (2015), values of all standardized (drought) indices time series were restricted to the interval [-3, 3] STD.

The *Gamma* distribution was chosen for the SPI for all variables (RhiresD, ERA5-Land), which is consistent with other studies and ~~WMO recommendations~~ recommendations of the World Meteorological Organization (WMO) (WMO and GWP, 2016; Stagge et al., 2015; Tschurr et al., 2020; von Matt et al., 2024). The SMRI was fitted by the *genlog* (*lnorm*) distribution for the snow-corrected precipitation series based on SPASS (ERA5-Land and OSHD). For the SPEI, the *genlog* distribution was found to perform best across time scales (see Fig. 4).~~Following Stagge et al. (2015), values of all standardized (drought) indices time series were restricted to the interval -3, 3STD.~~

SPI, SPEI and SMRI provide complementary (and standardized) information on hydroclimatic variability and drought-related processes facilitating integrated analyses of drought development and propagation and allowing consistent comparisons across catchments, regions and climates (Mwinjuma et al., 2026; Gebrechorkos et al., 2025; Tijdeman et al., 2022). Standardized indices are frequently employed in drought monitoring and early warning systems (DEWS; Tijdeman et al., 2020; Kchouk et al., 2022), drought propagation analysis or as proxy for various storage processes (Haslinger et al., 2014; Cammalleri et al., 2019; Raposo et al., 2023; Peña-Gallardo et al., 2019; Barker et al., 2016). Example use cases for drought event analysis and catchment response patterns are provided in sections A2 and A3 in Appendix A.

4.5 Climatology & Anomalies

Climatologies and anomalies are provided for all time series~~including the standard~~, including the time series of ~~extracted variables (see sections 3.1 and 3.2~~variables directly extracted from spatially gridded data products with no modifications except for spatial and temporal aggregation where required (see Sect. 4.1), derived indicators (~~Section~~ Sect. 4.2), standardized ~~(drought) indices (Section 4.4)~~drought indices (Sect. 4.4), and cumulative water deficits (~~Section~~ Sect. 4.3 and 4.6).

Both climatologies and anomalies are based on the reference period 1991–2020. The climatology is provided for two variants: i) using moving windows and ii) for fixed ~~periods~~time windows. The variants are available at the following time scales: daily (only i), monthly (both), seasonal (both), and annual (only ii). The moving window climatology was calculated by using a moving window of 31 days ($day-15$, $day=0$, $day+15$) for the monthly, a 3-month window (91 days) for the seasonal and a 6-month (183 days) window for the extended season time scale. The moving window climatology is calculated for DOYs 1–366 with NA-values set for February 29~~th~~ th in the case of non-leap years. Example time series for monthly anomalies in 2 m temperature, precipitation, evaporation, soil moisture, snow water equivalent, snowmelt and (potential) cumulative water deficits are shown in Fig. 6a-f,m-n.

The regular climatology is available for monthly, seasonal (DJF, MAM, JJA, SON), extended season (~~Mai–October~~May–October, November–March) and annual time scales. Using the moving window climatology, standardized anomalies have been derived by ~~calculating z-scores ($(value-\mu)/\sigma$)~~ first subtracting the climatological mean (μ) and then

dividing by the climatological standard deviation (σ) (also known as *z-scores*). The following climatological statistics are provided: minimum, maximum, mean, median, standard deviation, 5thth, 25thth, 75th th and 95th th percentiles. For the 7-day average streamflow series (M7Q) we also provide the 2ndnd, 10th th and 15th ~~percentiles~~ th percentiles which are frequently used in streamflow drought analysis and monitoring (see e.g., Van Loon, 2015; Stahl et al., 2020; Sarailidis et al., 2019; BAFU, 2025)
.

4.6 Cumulative streamflow deficits

Time series of cumulative streamflow deficits (CQD) were calculated based on negative streamflow anomalies (drought phases) by using the same procedure as for cumulative water deficits (see ~~Section~~ Sect. 4.3). CQD time series are provided for both fixed and variable threshold definitions. Fixed thresholds (e.g., a constant percentile threshold) are used for critical flow levels that do not change seasonally (e.g., directly linked to physcial/actual low-flow or water scarcity situations) whereas variable thresholds (e.g., seasonally varying percentiles) account for seasonality and changing flow regimes, allowing drought phases and deficits to be identified relative to expected (seasonal) conditions ("anomalies", Stahl et al., 2020; Van Loon, 2015; Brunner et al., 2019a; von Matt et al., 2024). Hence, variable threshold definitions are often used to analyse seasonally varying streamflow (drought) generating processes or to understand drought propagation mechanisms (Brunner et al., 2023, 2022; Hammond et al., 2022). For the fixed threshold definition, daily M7Q anomalies were derived for ~~the yearly Q347-threshold events (≈ the yearly~~ events exceeding the Q347 threshold, defined as the daily flow rate exceeded for 347 days per year (i.e., the 347-day exceedance flow, roughly corresponding to the 5th ~~percentile, see Section ??~~th streamflow percentile; see Sect. 5.3). For the variable threshold definition, daily M7Q anomalies were calculated for the following monthly (31 days) and seasonal (91 days) percentiles: 2ndnd, 5thth, 10thth, 15thth, 25thth, 50th th (median) and mean. Cumulative deficits are physically interpretable and in the case of cumulative water deficits [mm] and streamflow deficits [m^3/s] also physically comparable in terms of total runoff depth [mm]. Figure 6j shows CQDs for both fixed and variable threshold definitions for the year 2022 for catchment *2034 - Broye, (Payerne, Caserne d'aviation)*.

4.7 Identification of drought events

We define drought events as coherent phases of non-zero deficits for cumulative deficits (CWD, PCWD and CQD) and as negative M7Q-based streamflow anomalies for streamflow droughts. Streamflow drought phases were extracted for the same percentiles and time scales as used for CQDs (see ~~section~~ Sect. 4.6), namely for monthly (31 days) and seasonal (91 days) percentiles: 2ndnd, 5thth, 10thth, 15thth, 25thth, 50th th (median) and the mean. ~~For each event definition~~Streamflow events were also extracted for the fixed (yearly) Q347 threshold (see Sect. 4.6 and 5.3).

An event starts on the first day values fall below the threshold value and lasts until values exceed the threshold again (see Fig. 5). For each variant, the event time series consists of consecutively numbered event phases and information on the event duration since the start ~~.~~(i.e, an event with a duration of 5 days is represented in the time series as: "1 1 1 1 1" (event phase number), "1 2 3 4 5" (duration since start)). Additional event characteristics (e.g., lowest value during a phase) can easily be derived by the

user in combination with the indicator time series. A minor pooling ~~for~~ of hydrological drought events is introduced by using the 7-day average streamflow (M7Q) time series which merges closely successing and potentially dependent individual events to one single event as a result of the smoothing of large day-to-day fluctuations (Tallaksen and Van Lanen, 2004; Hisdal and Tallaksen, 2000; Tallaksen et al., 1997; Sarailidis et al., 2019). Streamflow drought events based on both fixed and variable threshold definitions were used for the event shadings in Fig. 6.

5 Catchment descriptors

Catchment descriptors were extracted from spatial datasets containing information on hydro-terrestrial characteristics (e.g., soil suitability maps), catchment (station) metadata (see ~~Section~~ Sect. 3.3) and the extracted hydro-meteorological time series (e.g., climatology; see ~~Section~~ Sect. 4.5). All catchment descriptors provide only static (time-invariant) catchment information. Catchment descriptors are provided as single-value catchment-level information. An example use case of catchment grouping/regionalization based on catchment descriptors is presented in Sect. A1 in Appendix A.

5.1 ~~Extraction of catchment~~ Field-based descriptors

Spatially non-overlapping polygon datasets (e.g., soil suitability maps) typically provide categorized values for variable-specific classes (e.g., soil depth classes are *shallow*, *medium*, *deep*, *very deep*). To extract catchment-level information, polygon-based information was first rasterized to a spatial grid identical to the MeteoSwiss spatial climate analyses grid products (in both extent and resolution). The rasterization was done by using the *rasterize* function of the *terra* R-package (Hijmans, 2023). Each grid cell only contains the value of the category with the largest overlap. ~~Slope and aspect values were derived from the swissALTI3D digital elevation model (DEM, see section 10 for a download link) by using the standard *terra* R-package function *terrain* (Hijmans, 2023). A custom categorization was then applied to the resulting grid values (e.g., for slope 0–30, 30–60, etc.).~~ The percentage overlap with the catchment area was then assessed for all variable-specific classes by using the *exact_extract* function (as for time series) and adjusting the aggregation function to fractions ("frac"; see Baston, 2023). Catchment area overlap fractions are provided for all categories. Descriptors with multiple classes can also be reduced to a single dominant category represented by the largest percentage overlap ("proportion"). An example is shown for the biogeographic regions in Fig. 7~~. However, the class~~ a. Reducing a specific catchment descriptor to one single (dominant) category (e.g., derived via largest overlap percentage) may however lead to a loss in explanatory power as the category with the largest overlap ~~does not necessarily correspond to~~ may not necessarily be the most representative ~~, as multiple categories can share similar proportions of the catchment area~~or most influential for streamflow (drought) analysis.

5.2 ~~Derivation of other catchment~~ Feature-based descriptors

~~Additional catchment descriptors were derived from the remaining descriptive input products (catchment metadata, karstic sources, catchment outlines and hydrography) and the calculated hydro-climatology (see Section 4.5).~~Two descriptive variables related to catchment shape and drainage were derived in R by using the catchment outlines, namely the *basin shape*

index (BSI) and *drainage density*. The HYD-RESPONSES dataset provides two BSI variants. The first variant is derived based on a ratio between area and basin length (A/L^2) known as form factor (Horton, 1932) and the second variant is based on a ratio between the catchment area and the area of the circle with the smallest radius encircling the entire catchment (A_{catch}/A_{circle}) known as circularity ratio (Miller, 1953). Both indices range from 0 to 1. Both are frequently used (also in combination) as morphometric catchment indicators (see e.g., Das et al., 2022; Pisupati and Ratnakar, 2025). Albeit providing similar information, the form factor is primarily controlled by basin length and hence provides information on catchment elongation, while the circularity ratio is more sensitive to basin shape accounting for complex/irregular shapes resulting in larger areas (for more information on basin shape indices see Das et al., 2022). ~~For more informationsee Das et al. (2022) .~~ The drainage density denotes the ratio between the catchment area and the total length of streamflow channels (both natural and stormwater drainage infrastructure; Dingman, 1978; USGS, 2023). The drainage density was calculated by using the swissTLM3D hydrography dataset (see ~~section~~ Sect. 10 for a download link). Both indices (BSI and drainage density) are frequently used in flood-related studies but may also provide valuable information during low-flow periods as high-intensity precipitation events are a relevant factor for (streamflow) drought recovery (Eekhout et al., 2018; Floriancic et al., 2022; Lee and Ajami, 2023; Matanó et al., 2024; Qiu et al., 2021; Tarasova et al., 2024; Vicente-Serrano et al., 2022; Wu et al., 2022; Xu et al., 2023). Further, also the overlap percentage with the Swiss territory (swissBOUNDARIES3D, see ~~section~~ Sect. 10 for a download link) is provided for each catchment and can be used to exclude catchments with significant portions outside of Switzerland which goes along with a limited coverage in both hydro-meteorological and catchment descriptor input datasets (see ~~Sections~~ Sect. 3.2 and 3.3) ~~.~~ for ca. 12.5 % of catchments (see Sect. 2). Information on karstic sources is provided as the number of sources per catchment and km^2.

5.3 Time series-based and climatological descriptors

Several indices related to streamflow characteristics (low flow, responsiveness, baseflow and flow stability) are provided in the HYD-RESPONSES dataset. The Q347 (Aschwanden, 1992; Aschwanden and Kan, 1999) is a low flow index used as the basis for water abstraction restrictions in Switzerland and corresponds to the ~~5^{th} streamflow percentile (low flows)~~ daily flow rate exceeded for 347 days per year. The Q347 was derived from the flow duration curve (FDC) ~~. The Q347 was derived~~ by using the *hydroTSM* R-package (Zambrano-Bigiarini, 2020) and corresponds roughly to the 5^{th} streamflow percentile (95^{th} percentile of 365 days $\approx$ 347, hence Q347). The baseflow index (BFI; Nathan and McMahon, 1990) is a widely used index linked to multiple catchment characteristics such as aquifer type, productivity and soil characteristics. The BFI provides information on the (base-)flow sustained during dry periods (e.g., by subsurface storages; Tallaksen and Van Lanen, 2004; Bloomfield et al., 2021; Van Loon and Laaha, 2015). The BFI was derived using the *baseflow* function of the *lfstat* R-package (Laaha and Koffler, 2022) and is shown in Fig. 7. Stoelzle et al. (2020) introduced the delayed-flow index (DFI) which breaks down the BFI into individual hydrograph components. The components include fast, intermediate, slow and base responses and potentially reflect various storage processes contributing to the overall streamflow response (e.g., snowmelt and groundwater). The DFI was derived by using the *delayedflow* R-package (https://modche.github.io/delayedflow/; see also Stoelzle et al., 2020). The last two indices related to streamflow behaviour are the "flashiness" or R-B-index (Baker et al., 2004) which represents the ratio of

the sum of day-to-day streamflow changes divided by the total streamflow and the flow-stability index which relates the mean annual minimum flows (MAM_q) to the mean annual flow (~~MAM~~/MQ; MAM_q/MQ). The remaining catchment descriptors were derived from the extracted hydro-meteorological time series and/or their respective climatology. Information on average precipitation, temperature, evaporation, snow water equivalent, streamflow, the fraction of precipitation falling as snow and the runoff fraction (Q/P) are provided ~~, partly on both monthly and yearly time scales~~ on yearly scales for identifying broad climatic (i.e., water balance) and physiographic controls on hydrological behavior. Finally, monthly Pardé coefficients (PCs) are provided which indicate the contribution of monthly mean streamflow to the annual mean streamflow.

6 ~~Three example use cases~~Discussion

6.1 Relevance and Applications

The HYD-RESPONSES dataset addresses fundamental challenges in hydrological drought analyses by compiling and harmonizing multiple data sources into a coherent catchment-scale framework, enabling multi-variable drought analyses in Switzerland. Drought (deficit) indices derived from two high-resolution snow climatologies for Switzerland (SPASS, OSHD) also allow for in-depth quantitative analyses on the contribution of snow processes to cross-seasonal drought propagation in Alpine catchments (Staudinger et al., 2014, 2017; Brunner et al., 2023). A combined used of standardized indices (SPI, SPEI, SMRI at 1—24 month scales) and non-standardized cumulative deficits (CWD, PCWD, CQD) facilitate multi-scale (drought) deficit and catchment response sensitivity assessments and allows for a concurrent anomaly-based and physically interpretable characterization of drought deficits (Raposo et al., 2023; Van Loon, 2015; Wu et al., 2020; Baez-Villanueva et al., 2024; Stocker et al., 2023). By providing time series for many relevant variables for drought monitoring (precipiation, temperature, evaporation, soil moisture and streamflow; see e.g., WMO and GWP, 2016) at daily temporal resolution, the HYD-RESPONSES dataset may also be used for training machine learning models such as Random Forests (RFs; e.g., Floriancic et al., 2022) or Long Short-Term Memory models (LSTMs; see Kratzert et al., 2018; Lees et al., 2022; Kratzert et al., 2023) which have recently emerged as promising approach for rainfall-runoff modeling (Kratzert et al., 2018, 2019; Lees et al., 2022). Three example applications of the HYD-RESPONSES dataset are illustrated in Sect. A1, A2 and A3 in Appendix A.

Although the dataset was developed for Switzerland, the methodological framework — combining in-situ observations, gridded products, and reanalysis into catchment-scale time series — is transferable, with requirements that scale with data availability. Replication requires four essential components: (1) streamflow observations with defined catchment boundaries, (2) meteorological forcing data (precipitation, temperature), (3) snow information for mountain regions, and (4) static catchment descriptors (e.g., information on soils, geology, topography). While the first component is often limiting, the second component is decisive for the applicability of the dataset on specific use cases. Switzerland can leverage from high-density observational station networks resulting in high-quality spatially gridded hydro-meteorological products (see e.g., MeteoSwiss, 2024). With known biases in mind (especially over complex terrain), ERA5-Land is a viable alternative

by providing temporally and physically consistent variables at sufficiently high spatial resolution for comparative catchment studies and machine learning applications aiming for generalizable results in regions where observational networks are less dense (Muñoz-Sabater et al., 2021; Dalla Torre et al., 2024; Scherrer et al., 2023). For local operational drought management or absolute deficit quantification, reliable high-resolution observational products remain however preferable. Several recent developments address observational data limitations, including the Caravan global community dataset (Kratzert et al., 2023), rapidly advancing machine learning-based bias correction methods for downscaling reanalysis products such as ERA5-Land (Menapace et al., 2025; Najafi et al., 2026; Zhang et al., 2025) or advances in developing high-quality remote sensing-based products for soil moisture (e.g. SMAP; see Brocca et al., 2024b; An et al., 2025), snow (e.g., ICESat-2 Besso et al., 2024), evaporation (see e.g., Anderson et al., 2024) and terrestrial water storage (e.g., GRACE-FO; Rodell and Reager, 2023).

6.2 Limitations and cautionary notes

The HYD-RESPONSES time series are provided for product-specific periods, and the spatial coverage is restricted to Swiss territory for most of the higher resolution MeteoSwiss and SLF products (TabsD, TminD, TmaxD, SPASS, SrelD, OSHD) as well as many catchment descriptor input datasets. Full coverage over the entire hydrological Switzerland is only available for ERA5-Land (all variables) and the MeteoSwiss RhiresD product (after 1992; see MeteoSwiss, 2021a). Catchments with significant areas outside of the Swiss national borders —approximately 12.5 % of the catchments (see Sect. 2) — may therefore be considered with caution or used solely for time series based on ERA5-Land variables only. Time series for standardized drought indices are provided only for the transformation variant based on the best-fitting distribution across all catchments to allow for catchment comparability (see e.g., Staudinger et al., 2014). The best-fitting distribution may however vary across catchments and climates (see e.g., Stagge et al., 2015). The HYD-RESPONSES dataset therefore also provides information on fits, missing values, and flags which can be used to exclude catchments with unsatisfying fitting and transformation properties from analyses. Field- and feature-based catchment descriptors were aggregated at catchment level via summarization (e.g., karstic sources) or percentage overlaps (see Sect. 5.1). Maximum percentage overlaps with catchment area may however only insufficiently account for spatial differentiation which could enhance the representation of factors most influential to streamflow evolution by accounting for spatial proximity to the stream/river courses (Tarasova et al., 2024; Floriancic et al., 2022).

Several known limitations are further related to the datasets used to compile the HYD-RESPONSES data. ERA5-Land is a state-of-the-art reanalysis product provided at a higher spatial resolution than the standard ERA5 reanalysis (Hersbach et al., 2020; Muñoz-Sabater et al., 2021). The higher spatial resolution results in a better depiction of soil moisture, lakes, river discharge estimations, and the orographic enhancement of precipitation (Muñoz-Sabater et al., 2021). ERA5 and ERA5-Land datasets however share most of the parameterizations, as ERA5-Land consists of output from the ECMWF land surface model driven by downscaled and elevation-corrected ERA5 data (Muñoz-Sabater et al., 2021). Despite the advantage of higher spatial resolution over ERA5, a grid resolution of 9 km still has limitations over complex high-altitude terrain. The extracted time series related to snow water equivalent should be used with caution, as snow depth in ERA5-Land is

of mixed quality depending on geographical location and altitude (Dalla Torre et al., 2024). Scherrer et al. (2023) showed that ERA5-Land overestimates SWE at high elevations with larger biases in the southern compared to the northern Alps. Higher-resolution datasets such as SPASS (Marty et al., 2025) and OSHD (Mott, 2023; Mott et al., 2023) should hence be preferred over ERA5-Land. Note however that all snow-related datasets have problems in representing small SWE amounts at low altitudes (Scherrer et al., 2023; Michel et al., 2023; Marty et al., 2025). Caution is also required when using the snow-corrected precipitation (water input) time series. The time series corrected by the ΔSWE series consider both snowfall ($\Delta SWE > 0$) and snowmelt ($\Delta SWE < 0$), while the correction based on snowmelt variables only accounts for snowmelt (smlt in ERA5-Land and romc in OSHD; see Table 1 and Tables B1 and B2 and Sect. 4.2.2). Snowmelt-corrected precipitation time series may therefore be of limited use during the main snow accumulation season but can still provide valuable information during the snowmelt season.

Another limitation of the ERA5-Land dataset is the parameterization of subgrid-scale processes and the representation of subsurface storages that affect evapotranspiration (e.g., fixed maximum storage volume assumption; see Muñoz-Sabater et al., 2021). Key processes such as dynamic groundwater–vegetation interactions, irrigation withdrawals, and adaptive rooting strategies are hence not represented and may lead to biases in evapotranspiration responses (Muñoz-Sabater et al., 2021; Dalla Torre et al., 2024; Wood et al., 2025; Stocker et al., 2023). Although ERA5-Land compares more favorably with in situ soil moisture and evapotranspiration observations than previous reanalyses (e.g., ERA5), considerable discrepancies remain, especially in dry summers and in regions with heterogeneous land cover (Scherrer et al., 2022; Fluhrer et al., 2025). Given these limitations, drought indicators based on ERA5-Land evapotranspiration should generally be interpreted with caution. This limitation is further compounded by the fact that the validation of long-term soil moisture and evapotranspiration remains challenging due to the scarcity of consistent observational datasets, particularly at high spatial resolution and over multi-decadal time scales (Hirschi et al., 2020; Yi et al., 2024; Mukherjee et al., 2018). Many state-of-the-art evapotranspiration products are limited in temporal or spatial extent and can be affected by gaps and cloud contamination (e.g., remote sensing-based products; see Yi et al., 2024). ERA5-Land thus remains one of the few datasets providing spatially consistent and continuous long-term evapotranspiration estimates with sufficiently high spatial resolution over Switzerland.

Additional caution is warranted when using HYD-RESPONSES water balance time series (and indicators derived from them) when they were derived by combining ERA5-Land evapotranspiration with (snow-corrected) precipitation from independent data sources (RhiresD, OSHD, and SPASS). While ERA5-Land variables are internally consistent, the combination with independent data sources may lead to systematic biases in absolute deficit estimates. This limits the interpretability of absolute cumulative deficits but does not invalidate the approach for comparative, process-oriented drought analyses across regions and catchments (e.g., drought propagation, catchment response sensitivities). Relative measures of cumulative deficits, their temporal evolution, and their normalization through ratios (e.g., CWD/PCWD)

can still provide valuable insights, even when absolute magnitudes are uncertain. In such contexts, relative anomalies, temporal evolution, and spatial patterns are more informative than absolute deficit magnitudes. Studies have further demonstrated coherent representation of major drought events (e.g., drought years 2003 and 2018) across datasets, which supports the usability of combined indicator time series when known limitations are adequately taken into account (Scherrer et al., 2022; Wood et al., 2025). Note that the HYD-RESPONSES dataset also provides complementary water balance and SPEI time series derived from ERA5-Land variables only, providing consistent metrics and opportunity for comparisons among data products. Guidance on the usage and reliability of all HYD-RESPONSES time series products is provided by a classification based on three reliability levels (see Sect. 4 and Table 3). The levels are based on the origin of the underlying data, the extent to which variables rely on (model) assumptions, and the degree of processing applied to derive the hydro-meteorological time series.

7 Complementary datasets

Complementary datasets provide a wide range of additional catchment descriptors and hydro-meteorological time series. An overview of datasets and variables is provided in Table 4. The FOEN provides additional geodata related to both surface and groundwater via the Hydrological Service (https://www.bafu.admin.ch/bafu/de/home/themen/wasser/zustand/karten/geodaten.html). The datasets include additional catchment descriptors with information on population density, catchment areas covered by forest and agriculture (among others) as well as information on water quality aspects and sewage. The FOEN further operates both a groundwater monitoring network (NAQUA) providing continuous groundwater measurements for selected point locations (BAFU, 2019) and a water quality measurement network (NAWA) providing information on concentration and loads of important dissolved compounds (e.g., pH, electric conductivity, nutrient contents; BAFU, 2023).

The "Catchment Attributes and Meteorology for Large-sample catchment Studies" (CAMELS) datasets aim at providing a consistent set of hydro-meteorological time series and catchment descriptors over a large sample of hydrological catchments on country level (Clerc-Schwarzenbach et al., 2024). The catchments in the Swiss version of the CAMELS data (CAMELS-CH; Höge et al., 2023a) are largely congruent with our dataset. The only exception is station 2646, which is only contained in the HYD-RESPONSES dataset. Note that the HYD-RESPONSES dataset provides only a sample subset of 184 catchments. The CAMELS-CH dataset provides valuable complementary catchment-level information on glacier changes (based on GLAMOS, for details see Höge et al., 2023a), land use, hydro-geological and hydro-terrestrial information (e.g., the contributions of various grain size categories and bulk-density) as well as anthropogenic disturbances (e.g., hydropower and reservoir capacities). CAMELS-CH further provides modelled time series based on the hydrological model PREVAH (see e.g., Höge et al., 2023a; Viviroli et al., 2009). The CAMELS-CH dataset is freely available from Zenodo (https://zenodo.org/records/10354485; Höge et al., 2023b).

The CombiPrecip dataset (MeteoSwiss) provides high-resolution (10 minutes, 1×1 km) precipitation fields derived from a combination of radar and station measurement data (Sideris et al., 2014). The CombiPrecip dataset could be a valuable addition

Table 4. Datasets compatible and complementary to the HYD-RESPONSES dataset.

Dataset	Short description	Provider
Accompanying catchment information	Includes catchment proportions of forests, agricultural (crop) land, population, built-up area and more	FOEN
Groundwater measurement network (NAQUA)	Groundwater measurements	FOEN
Water quality measurement network (NAWA)	Information on water quality parameters	FOEN
CAMELS-CH (Höge et al., 2023b)	Swiss version of the Catchment Attributes and Meteorology for Large sample catchment Studies (CAMELS) dataset	via Zenodo
MeteoSwiss CombiPrecip (CPC)	High-resolution precipitation fields at ground based on a combination of radar and measurement data	MeteoSwiss
HydCHeck (Streeb et al., 2024)	Detailed evaluation of influences and disturbances of the streamflow at NAWA measurement stations	FOEN

for studying drought recovery where extreme precipitation is often considered an important factor (Wu et al., 2022).
The HydCHeck project (Streeb et al., 2024) evaluated the influence of (anthropogenic) disturbance factors on streamflow at stations of the National Surface Water Quality (NAWA) Programme (BAFU, 2023). The evaluated NAWA stations are largely (87.5% of the stations) congruent with the HYD-RESPONSES dataset. The HydCHeck dataset provides catchment-level information on the magnitudes for all evaluated disturbance categories including water storage and regulation, hydropower, sewage water, constructions, agriculture as well as drinking and groundwater. The overall impact on several hydrological properties including low-, mid- and high-flow regimes as well as short-term effects and hydraulics is provided as categorical information (from "not disturbed" to "strongly disturbed"). For more information see Streeb et al. (2024).
As part of the planned Swiss National drought early warning system (DEWS), both a high-resolution remote-sensing based evaporation product (V. Humphrey, pers. comm.) and an automatic soil moisture measurement network are under development at MeteoSwiss, ETH Zurich and WSL and may become a valuable addition in a future.

8 Conclusions

The HYD-RESPONSES dataset contains data for 184 Swiss catchments that cover a variety of streamflow regimes, mean altitudes, catchment areas, and anthropogenic influences/disturbances. The catchments cover all biogeographic regions of Switzerland. The HYD-RESPONSES dataset provides daily streamflow data and daily hydro-meteorological time series extracted from gridded data products of MeteoSwiss (TabsD, RhiresD, TmaxD, TminD, SrelD), Meteoswiss and SLF (SPASS),

SLF (OSHD) and ECMWF (ERA5-Land). The variables include temperature, precipitation, evaporation, sunshine duration, solar radiation, snowmelt, snow water equivalent, soil moisture, surface runoff, runoff, and streamflow. HYD-RESPONSES further provides derived variables related to streamflow (e.g., M7Q), water balance (e.g., P–E) and snowfall. Additionally, three standardized drought indices (SPI, SPEI, SMRI) for accumulation windows from 1 to 24 months and information on the (non-standardized) cumulative water deficit (CWD), the potential cumulative water deficit (PCWD) and cumulative streamflow deficit (CQD) are provided.

The dataset also provides information on (streamflow) drought events (occurrence and duration). For each catchment, the drought events have been identified based on fixed and on seasonally varying percentile thresholds.

The combination of data sources, the information on hydro-meteorological variables (mainly temperature, precipitation and snow), the derived indices (water balance, cumulative water deficits, standardized drought indices, climatology and anomalies) allow for a multi-purpose use and various analytical approaches such as time series analysis (e.g., Kratzert et al., 2018; Lees et al., 2022), drought propagation and catchment sensitivity analysis (e.g., based on principal component analysis and clustering; Jehn et al., 2020) and changes in rainfall-runoff relationships during hydrological droughts (e.g., Wu et al., 2021).

The HYD-RESPONSES dataset can easily be combined with complementary datasets such as CAMELS-CH (Höge et al., 2023a) and HydCHeck (Streeb et al., 2024). The catchment time series vary in length (subject to station initialization), the hydrological time series are provided for the entire measurement period along with information on data homogeneity (see BAFU (2024) for more details).

Limitations exist for catchments extending beyond the Swiss borders. The catchment descriptors were extracted from datasets mainly covering Swiss national territory. The MeteoSwiss-based datasets cover only Switzerland except for RhiresD, which covers the entire hydrological Switzerland from 1992 onward. In summary, the dataset provides a state-of-the-art data basis to study droughts in Switzerland.

9 Code and data availability

The HYD-RESPONSES dataset is freely available (CC BY 4.0) from Zenodo (https://doi.org/10.5281/zenodo.15748821; von Matt et al., 2026). Regular updates are not planned. An R tutorial on how to use and combine the different data products is provided with the dataset but can also be accessed on GitHub (https://github.com/codicolus/HYD-RESPONSES).

As of now, MeteoSwiss gridded spatial analyses products (MeteoSwiss, 2021a, b, c) are not available for free but will be available for free in the course of 2025 (MeteoSwiss, 2025). The preliminary snow climatology for Switzerland (SPASS; see Michel et al., 2023; Marty et al., 2025) was provided directly by MeteoSwiss and is not yet available for public use. The SLF snow climatology (OSHD; Mott, 2023; Mott et al., 2023) was published under the WSL Data Policy and can be downloaded via Envidat (https://www.envidat.ch/#/metadata/climatological-snow-data-1998-2022-oshd). The hourly ERA5-Land dataset (Muñoz-Sabater et al., 2021) is accessible via the Copernicus Climate Data Store (CDS) (see https://cds.climate.copernicus.eu/datasets/reanalysis-era5-land?tab=download). Daily streamflow time series can be requested

via the Hydrological Service of the FOEN via https://www.bafu.admin.ch/bafu/de/home/themen/wasser/zustand/daten/messwerte-zum-thema-wasser-beziehen.html. The soil suitability maps (FOAG), the hydrogeological map (FOEN) and the lithological map (Swisstopo) are available from https://opendata.swiss or directly via Swisstopo (https://www.swisstopo.admin.ch/de/geokarten-500-vektor). Directly available from Swisstopo are also the datasets swissTLM3D Hydrography (https://www.swisstopo.admin.ch/de/landschaftsmodell-swisstlm3d#swissTLM3D---Download) and swissBOUNDARIES3D (https://www.swisstopo.admin.ch/de/landschaftsmodell-swissboundaries3d). Further available via https://opendata.swiss are the Biogeographic regions (https://opendata.swiss/de/dataset/biogeographische-regionen-der-schweiz-ch; see also BAFU (Eds.), 2022) and information on karstic springs and swallow holes (also produced by the FOEN; https://opendata.swiss/de/dataset/quellen-und-schwinden-in-karstgebieten). Data used for the overview map of the study region (Fig. 1) is available for free from Swisstopo and FOEN. Datasets used include: the digital height model DHM25 (https://www.swisstopo.admin.ch/de/hoehenmodell-dhm25) and the general hydrological background map (downloadable via https://opendata.swiss; see https://opendata.swiss/en/dataset/generalisierte-hintergrundkarte-zur-darstellung-hydrologischer-daten).

The software used to compile the datasets are all open-source and contain the following R-packages available via CRAN: *tidyverse* (https://cran.r-project.org/web/packages/tidyverse/index.html; Wickham et al., 2019), *exactextractr* (https://cran.r-project.org/web/packages/exactextractr/index.html; Baston, 2023), *sf* (https://cran.r-project.org/web/packages/sf/index.html; Pebesma, 2018), *lfstat* (https://cran.r-project.org/web/packages/lfstat/index.html; Laaha and Koffler, 2022), *SCI* (https://cran.r-project.org/web/packages/SCI/index.html; Gudmundsson and Stagge, 2016; Stagge et al., 2015) and *stars* (https://cran.r-project.org/web/packages/stars/index.html; Pebesma and Bivand, 2023).
Available via Github are the R-packages *cwd* (Stocker (2021); available via: https://github.com/stineb/cwd), and *delayedflow* (Stoelzle et al. (2020); available via: https://modche.github.io/delayedflow/).

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Appendix A: Exemplary use cases

The different data types can be combined to comprehensively analyse hydrological streamflow droughts in response to various hydro-meteorological indicators. This section presents three use cases: catchment regionalization, in-depth event analysis, and composite analysis. A comprehensive R-tutorial on how to read and combine the different data products is provided with the dataset but can also be accessed via Github (https://github.com/codicolus/HYD-RESPONSES).

A1 Catchment grouping

For some applications, catchments need to be grouped by similarity, as measured by a set ~~of~~ catchment descriptors related to hydro-meteorological, terrestrial characteristics and/or anthropogenic ~~catchment descriptors~~ disturbances (e.g., Tarasova et al., 2024). As an example application, we show the distribution of catchment coverage fractions across biogeographic regions for the soil characteristics *soil depth, skeletal content, water logging, permeability, and water storage capacity* (Figure A1).

The percentage coverage distributions reveal notable differences in soil characteristics and their subcategories. Catchments in the Swiss Plateau region are characterized by larger coverages of deep to very deep soils with a mostly poor to medium skeletal content, normal soil permeability and good water storage capacities (see Fig. A1). Alpine catchments, on the other hand, are characterized by shallower soils (especially the Southern Alps) and a higher skeletal content. Soils in the Alps further have almost no water logging and a low water storage capacity. Soils with a (weakly) inhibited permeability or with a very good water storage capacity are infrequent across all biogeographic regions. ~~An other example for catchment grouping is the~~

Whereas soil characteristics reflect differences in terrestrial catchment characteristics, the streamflow regime types are more indicative of hydro-climatic streamflow generation processes (see Sect. 2, Fig. A2). In Alpine environments, snow- and glacier processes are important contributors to streamflow generation. Hence, both Western and Eastern Central Alps show predominantly glacial and nival streamflow regime types of which 86 % are shared among regions (see Fig. A2a,b). In contrast to the Central Alps, the hydro-climatology of the Southern Alps is more Mediterranean with the Alps often described as "Alpine divide" between the warmer and drier conditions in the South and more temperate Atlantic influences in the North (NCCS, 2025; Haslinger et al., 2019). Despite similar streamflow generating processes, the streamflow regime types in the Southern Alps are differentiated from their counterparts in the Central Alps (*méridional*, see Fig. A2d) and, in contrast to the Central Alps, can also include stronger contributions of pluvial processes. The Northern Pre-Alps show the largest variability of streamflow regime ~~type classification for Switzerland (see e.g., Aschwanden and Weingartner, 1985; Weingartner and Schwanbeck, 2020). Figure A1 in Appendix A shows the incidence of streamflow regime~~ types ~~across biogeographic regions~~(Fig. A2c). As transitional region between Alps and lowlands, streamflow regime types also show a large variety of processes contributing to streamflow generation including glacial, nival and pluvial processes. The Jura is a subalpine mountain range characterized by a strongly karstified geology. The variability of streamflow regime types is the lowest among biogeographic regions and consists mainly of locally specific types (*jurassien*) influenced by a mixture of nival and pluvial processes (Fig. A2e). The Swiss Plateau region is equivalent to the lowlands and hence is dominated by pluvial streamflow regime types (Fig. A2f).

Note that whereas catchments smaller than 500 km^2 allow for a more distinct discrimination in streamflow regime types and hence streamflow generating processes, catchments with an area larger than 500 km^2 are influenced by a multitude of streamflow generating and/or storage processes (e.g., groundwater contributions) and are hence provided as *mixed regime (>500 km^2)* type (see Sect. 2). The mixed regime type is the most prevalent among catchments (see Fig. 2 and occurs across all biogeographic regions (Fig. A2) being the most frequent regime type in Western and Eastern Central Alps as well as in the Northern Pre-Alps.

The catchment grouping to biogeographic regions presented here is again based on maximum overlap of catchment area with the specific biogeographic region (see also Fig. 7a). Biogeographic regions are frequently used for grouping catchments into groups of similar streamflow (generation) characteristics and provides usable results (see e.g., Brunner et al., 2019b; Muelchi et al., 2021b; von Matt et al., 2024). In specific cases where a categorization into biogeographic regions may not be unambiguous (see Sect. 5.1), categorization may be reconsidered by the user via alternative grouping (e.g. directly on streamflow regime types), alternative categorization approach (other than maximum overlap) or expert judgement. As such the streamflow regime types *nivo-pluvial jurassien* and *pluvial jurassien* could for example be recategorized to the Jura region (see Fig. A2f).

A2 Detailed Event analysis

The combination of hydro-meteorological indicators, standardized (drought) indices (SPI, SPEI, SMRI), ~~cumulative~~ (potential) ~~water (balance)~~ cumulative water and streamflow deficits (CWD, PCWD, CQD) and accompanying climatological anomalies allow for a detailed analysis of specific (streamflow) drought events. Drought-generating processes vary across catchments depending on hydro-climatological and terrestrial catchment characteristics, the season as well as on anthropogenic disturbances (e.g., Brunner et al., 2022; Van Loon and Van Lanen, 2012; Van Loon, 2015; Apurv et al., 2017). Except for glacier melt and groundwater, the HYD-RESPONSES dataset provides time series for all relevant hydro-meteorological indicators required to analyse (streamflow) drought generation, drought propagation as well as drought type classification.

Figure 6 illustrates time series for the year 2022 of a subset of relevant hydro-meteorological variables for catchment *2034 - Broye, Payerne, Caserne d'aviation*. This catchment is located in the western Swiss Plateau region (highlighted in Fig. A4). The year 2022 was an exceptional year with unprecedented combined heat and drought conditions over Europe (Tripathy and Mishra, 2023). The Broye catchment experienced low-flow conditions beyond a 100-year return period (BAFU (Eds.), 2023). In the Broye catchment, the lowest 7-day average streamflow values were observed between July and August (see Fig. 6i) Several streamflow drought events were identified for both yearly fixed (purple shading) and variable (green shading) threshold definitions ~~.~~(see Fig. 6 (all panels)). The longest events occur during the annual low-flow season for both definitions.

The year 2022 was also one of the warmest years on record with three heatwaves occurring in mid-June, mid-July and in the beginning of August (Imfeld et al., 2022). During the longest streamflow drought event in July 2022 (~~M7Q row in Figure~~ see Fig. 6i), evaporation anomalies begin to decline and become negative towards the end of the event (~~ET row in~~ see Fig. 6c). Concurrent strong negative soil moisture anomalies at shallow and deeper levels (see Fig. 6d) suggest that the successively decreasing evaporation anomalies may be related to increasingly depleted soil moisture storages resulting in limited

water availability for evaporation. Interactions between (subsurface) storage processes are however complex and also include groundwater–soil moisture interactions (e.g., Orth and Destouni, 2018).

~~Hydro-meteorological time series for the Swiss Plateau catchment 2034 - Broye, Payerne (Caserne d'aviation) for the year 2022. Color shadings in all panels highlight drought periods based on two definitions: yearly Q347 (pink, fixed threshold approach) and a moving monthly 15^{th} percentile threshold (green, variable threshold approach). **(a)** Moving monthly anomalies of the 2 m-temperature (T2m), positive anomalies are shown in red and negative anomalies in blue. **(b)** Moving monthly anomalies of the precipitation (P, RhiresD) **(c)** Moving monthly anomalies of the evaporation (ET, ERA5-land). **(d)** Moving monthly anomalies of the soil moisture volume (ESM ERA5-land), soil moisture anomalies are depicted for a near-surface SM-level (black, 0–7 cm) and the deepest level (blue, 100–289 cm) available from ERA5-Land. **(e)** Moving monthly anomalies of the snow water equivalent (SWE SPASS). **(f)** Moving monthly anomalies of the snowmelt (smlt, SPASS). **(g)** SPI colored by aggregation scales from 1- to 24-months. **(h)** SMRI colored by aggregation scales from 1- to 24-months. **(i)** Seven day average streamflow (M7Q). **(j)** The CQD time series shows the corresponding accumulated M7Q-deficits for both the fixed threshold approach (black) and the variable threshold approach (blue). **(k)** Absolute cumulative water deficit (CWD). **(l)** Potential cumulative water deficit (PCWD). **(m)** Monthly anomalies of the CWD (CWD anomaly). **(n)** Monthly anomalies of the PCWD. Time series of the cumulative water deficits for both absolute values and monthly anomalies are shown for both standard (black, P–E (P_e)) and snowmelt-corrected (blue, P–E+ΔSWE (PsmltSPASS_e)) variants. The same is shown for cumulative potential water deficits which are based on the potential water balance (P–PET (P_pev) and P–PET+ΔSWE (PsmltSPASS_pev)).~~

The HYD-RESPONSES dataset further provides information on cumulative (atmospheric) water deficits represented by standardized and non-standardized (drought) indices. For the Broye catchment, the 2022 streamflow drought events identified with the variable threshold (green shading) correlate well with shorter aggregation scales (1- to 3-monthly) SPI and SMRI indices in spring and summer. The correspondence between short-term precipitation deficits and streamflow droughts is, however, not consistent throughout the year. During the variable threshold streamflow droughts in mid-March to April, both SMRI-1 and SMRI-3 reach more negative values than their SPI equivalents, which ~~suggests a contribution of lacking snowmelt~~ indicates that lacking snowmelt contributed to the streamflow drought generation (see Fig. 6g,h). Lacking snowmelt as contributing factor is further confirmed by considering the larger and rather persistent negative anomalies in both SWE and snowmelt in the preceeding 1 to 3 months (see Fig. 6e,f).

Cumulative deficits in actual (CWD, Fig. 6k) and potential (PCWD, Fig.6l) water balance as well as streamflow (CQD, Fig. 6j) provide complementary information to the SPI, SMRI and SPEI in the form of non-standardized and hence physically interpretable deficit amounts. Cumulative streamflow deficits (CQD) show only two phases without deficit compensation for both drought definitions (Fig.6j). A shorter CQD phase coincides with the drought events in spring (variable threshold) and the shorter drought event in June (fixed threshold), while a longer phase coincides with the remaining shorter and longer streamflow drought phases in July and August before CQD is compensated by September 2022. For both CQD phases, the CQD is larger for streamflow droughts based on a variable threshold definition. Above average precipitation (+130 %; BAFU (Eds.), 2023) was reported in September 2022 and corresponds well with the compensation of CQD and is also reflected in the

positive monthly (31d) precipitation anomalies (P anomaly, Fig. 6b). Similar to the longest streamflow drought phases, also the largest ~~deficits in (actual)~~ cumulative deficits in water balance (CWD) occurred between May–August ~~2022. Larger CWDs during the warm season are consistent with the~~ 2022 (in terms of both absolute deficits and anomalies, see Fig.6k,m). Based on the seasonal climatology of both temperature and evaporation ~~with the~~ (highest values during summer), larger absolute CWDs are generally expected to occur during the warm season (not shown). Major CWD phases match streamflow drought phases remarkably well, especially for the variable threshold definition with one exception in April. The two longer streamflow drought phases in May–June further show the benefits of considering anomalies in the CWDs. While absolute CWDs were not compensated in between the streamflow droughts, the CWD anomalies indicate that the deficits returned to seasonal norm values (see Fig. 6m). Cumulative deficits in potential water balance (PCWDs, Fig. 6l) are more similar to cumulative streamflow deficits ~~(CQD)~~ for the variable-threshold definition ~~.~~ (CQD, Fig. 6j (blue line)). This reflects the different nature of CWDs and PCWDs. ~~The~~ Deficits based on the actual water balance ~~is~~ (P-E) are more strongly tied to the actual water availability and hence the individual streamflow ~~phases.~~ (drought) phases (Fig. 6i,k). The potential water balance (P-PET), on the other hand, represents the deficit that would have been accumulated under unlimited water availability. Similar to PCWD, CQDs reflect the integrated streamflow deficit over time while an actual deficit in terms of low streamflow levels does not necessarily have to exist (anymore).

A3 Composite analysis (catchment response patterns)

Composite analysis is a frequently used approach to understand the driving processes of a phenomenon such as droughts (see e.g., Bevacqua et al., 2021; Floriancic et al., 2020; Mahto and Mishra, 2024). By considering the median values of drought indicators across all streamflow drought events in a catchment, typical response patterns may become more evident and may allow for more generalized inferences on typical streamflow drought response patterns e.g., to precipitation deficits accumulated over various aggregation time-scales. Here, we present a composite analysis of median SPI values associated with streamflow droughts defined by the monthly 15^{th}-percentiles of the streamflow which corresponds to the highest of the low-flow percentile used for the Swiss national drought platform (see BAFU, 2025).

Note that streamflow drought characteristics and drought propagation processes may differ among catchments depending on hydro-meteorological climatologies, geological and terrestrial characteristics (e.g., aquifer, rock type, (soil) water storage capacity), seasonality of and differences in contributing streamflow (drought) generating processes and human disturbances (e.g., Van Loon and Laaha, 2015; Floriancic et al., 2022; Jehn et al., 2020; Apurv and Cai, 2020; Savelli et al., 2022; Haile et al., 2020; Brunner et al., 2022, 2021, 2023; Tijdeman et al., 2022; de Jager et al., 2022). We therefore separate the streamflow droughts and catchments by seasons winter (DJF, December–February), spring (MAM, March–May), summer (JJA, June–August) and autumn (SON, September–November) and by streamflow regime types. Six streamflow regime types are selected to capture a variety in dominant streamflow (drought) generating processes. These include glacial (*a-glaciaire, nivo-glaciaire*), nival (*nival alpin, nival méridional*) and pluvial (*pluvial jurassien, pluvial supérieur*) processes. The importance of precipitation deficits across scales is assessed using SPIs (SPI-1 to SPI-24). Streamflow drought events are only considered for the longest common homogeneous period across catchments (1991–2022). The selection is further restricted to catchments with

at least 10 streamflow drought events in each season (over the entire time series length) with a minimum duration of at least 10 days to enhance robustness and exclude minor droughts.

Median SPI values are mostly negative across all aggregation time-scales indicating that precipitation conditions co-occuring with streamflow droughts tend to be drier than normal. Several streamflow regimetype-specific response patterns are evident and change across seasons along with contributing streamflow (drought) generating processes.

Glacier melt is the dominant factor for the *a-glaciaire* regime type (Fig A3a–d). Streamflow levels are typically lowest in winter (January–March) as a result of precipitation falling as snow (intermediate storage) and highest in summer due to large contributions of glacier melt (Aschwanden and Weingartner, 1985; Weingartner and Schwanbeck, 2020; Muelchi et al., 2021b). Streamflow droughts of strongly glaciated catchments are not associated with moderate drought conditions at any SPI scale. In glacial and nival catchments a shift towards short-term precipitation deficits (SPI-1 to SPI-6) being associated with droughts is present across seasons and drought-generating processes. The transition towards shorter deficit scales emerges in summer for nival regime types (Fig. A3g,k) and in autumn for glacial regime types .(Fig. A3d). In pluvial and transitional regime types short-term precipitation deficits (mostly 1- to 3 months) are relevant throughout the year .(Fig. A3q–x). Seasonal shifts are also observed for pluvial and transitional regime types with mid- and long-term precipitation deficits becoming more relevant in summer and autumn .(Fig. A3t,w–x).

In addition to 3-monthly precipitation deficits, also mid- and long-term deficits become relevant in summer and autumn for (nivo-pluvial) catchments in the Jura region and catchments of the regime type *pluvial inférieur*. Compound moderate droughts are mainly observed for sub-yearly (1- to 9-monthly) scales with most extreme conditions on a 6-monthly scale in the Jura region (especially for nivo-pluvial catchments) and on a (6- to) 9-monthly scale for catchments of the regime type *pluvial inférieur*. In southern Switzerland, precipitation deficits tend to be relevant on longer scales compared to similar regime types north of the Alps. In contrast to nival catchments north of the Alps (*nival alpin*, Fig. A3i–l), droughts in the nival catchments south of the Alps (*nival méridional*, Fig. A3m–p) are associated with substantial precipitation deficits at longer aggregation times (9–24 months). The deficits occur in winter and in summer, but conditions are more extreme in summer (SPI $\approx$ -1.5) on scales longer than 15 months. Further, also 3-monthly precipitation deficits appear to be relevant for streamflow (drought) generation in summer (moderate drought conditions). In spring and autumn, mid- to short-term accumulation scales are more relevant. Interpretations of the differences between the south and north sides of the Alps should however be considered with caution due to the small catchment sample sizes and the spatial proximity of the two *nival méridional* catchments. The observed response patterns may therefore not be representative of nival catchments south of the Alps in general.

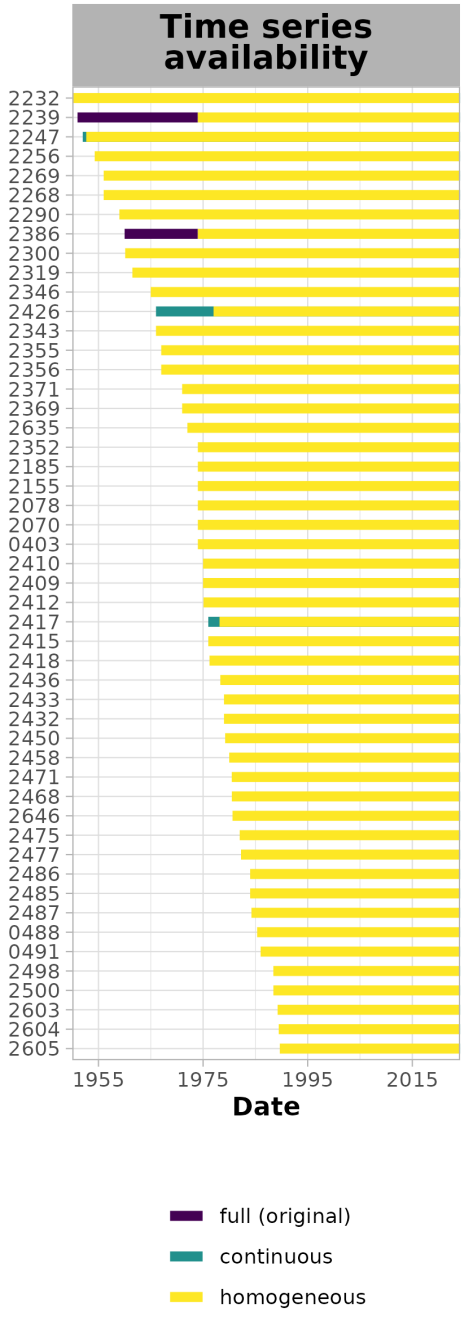

Figure 3. Streamflow time series availability for 50 example catchments. The colours indicate the periods covered by availability type. ~~Complete~~ Full is equivalent to the original time series provided by the FOEN. Continuous denotes the gap-checked time series and the homogeneous period accounts for homogeneity (starting at a breakpoint). In the case of overlapping periods, only the most important period type for analysis (e.g., homogeneous) is displayed. The importance of the periods for analysis is defined as follows: *homogeneous* is more important than *continuous* is more important than *full (original)*.

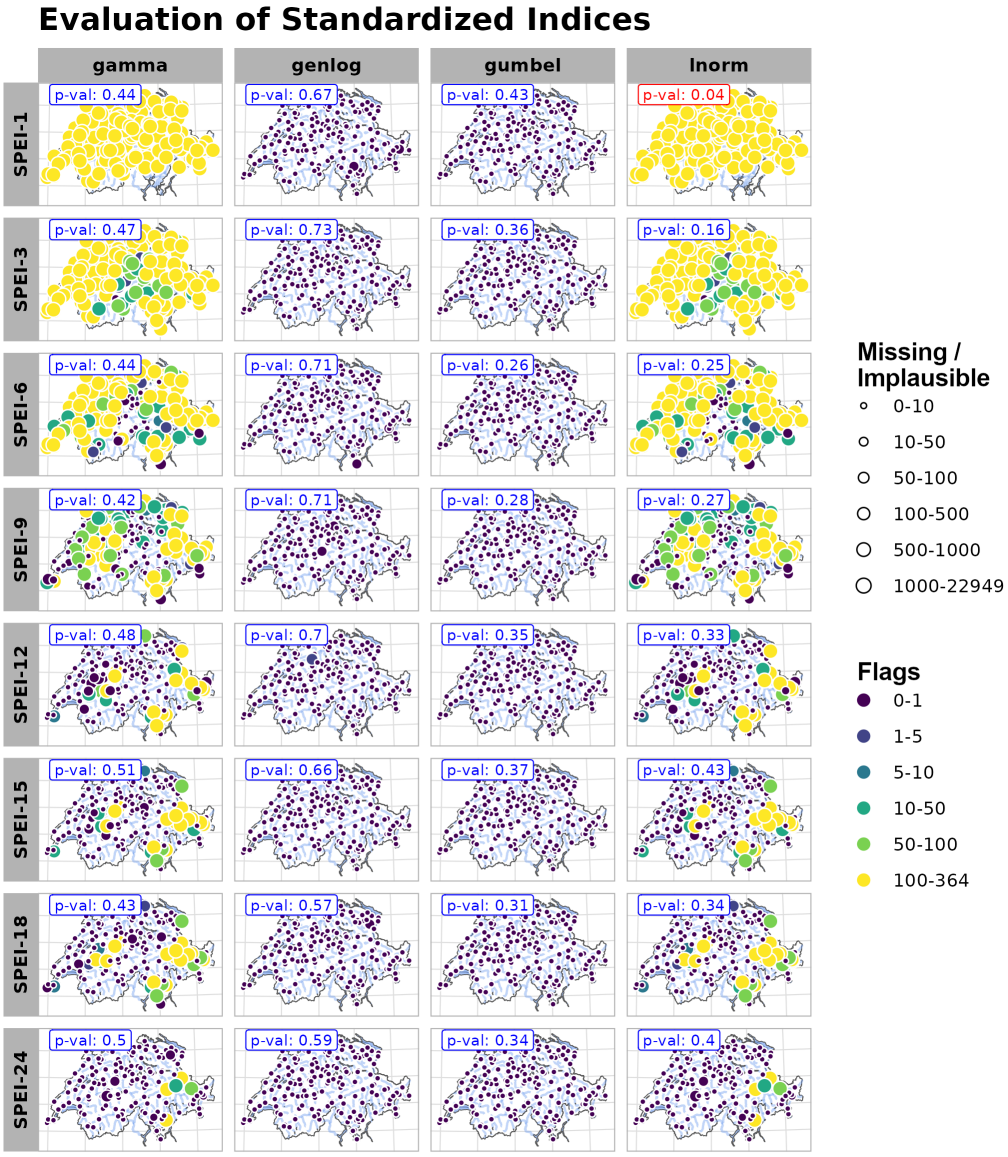

Figure 4. Evaluation statistics for the transformation of standardized (drought) indices. Information on the normality tests (p-values), flags and implausible/missing values (SPEI $\notin$ [-3,3]) for four example candidate distributions for the Standardized Precipitation and Evaporation Index (SPEI; Vicente-Serrano et al., 2010). The circle size indicates the number of missing and implausible values Colours show the number of flags (= convergence issues) returned by the fitting function of the *SCI* R-package (Stagge et al., 2015; Gudmundsson and Stagge, 2016) for all days of the year (DOY). The maximum number of flags is equivalent to 366. Median p-values of the Shapiro-Wilks normality test (Shapiro and Wilk, 1965) were calculated by considering all catchments and are coloured in red in case of rejection ($p < 0.05$). The final HYD-RESPONSES dataset only provides SPEIs fitted by the *genlog*-distribution (best choice based on the evaluation criteria).

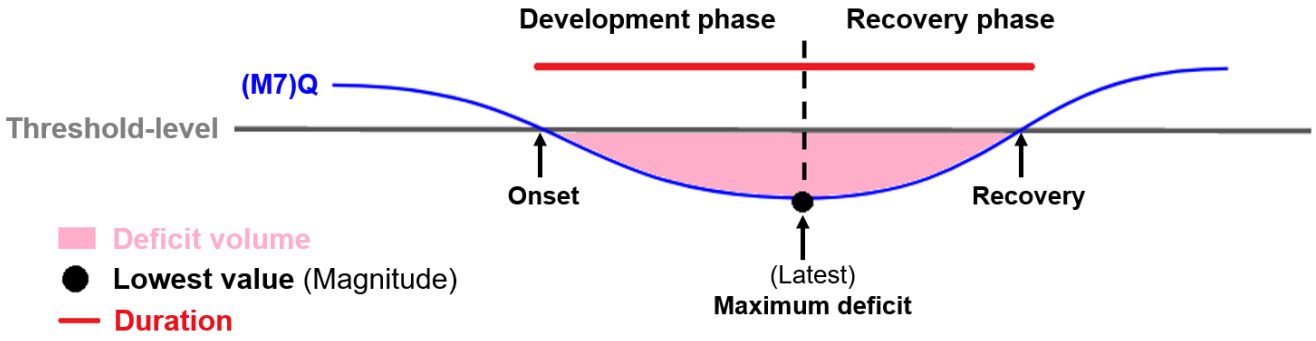

Figure 5. Schematic depiction of the event definition and phase subdivision. The extracted (streamflow) drought phases are characterized by duration, event start (onset), the latest date of the maximum streamflow deficit (anomaly), and event recovery. Additional characteristics are the drought intensity (deficit volume or accumulated deficit) and severity/magnitude (maximum streamflow deficit). The computation of other characteristics is left to the user.

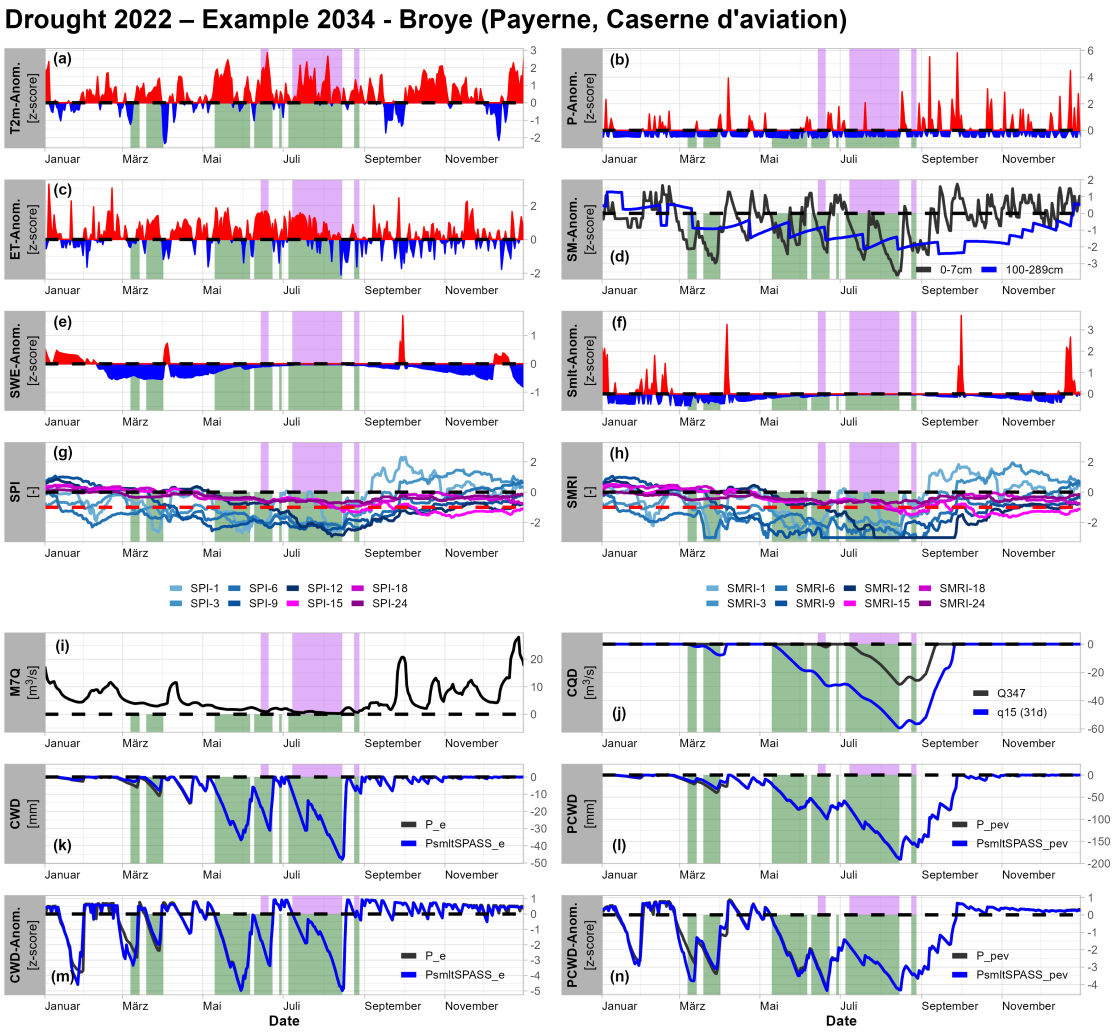

Figure 6. ~~Schematic depiction of~~ Hydro-meteorological time series for the ~~event definition and phase subdivision. The extracted~~ Swiss Plateau catchment 2034 - Broye, Payerne (~~streamflow~~Caserne d'aviation) for the year 2022. Color shadings in all panels highlight streamflow drought ~~phases are characterized by duration~~events for two definitions: yearly Q347 (pink, ~~event start~~ fixed threshold approach) and a moving monthly 15th percentile threshold (~~onset~~green, variable threshold approach). **(a)** Moving monthly anomalies of the 2 m-temperature (T2m), positive anomalies are shown in red and negative anomalies in blue. **(b)** Moving monthly anomalies of the ~~latest date~~ precipitation (P, RhiresD) **(c)** Moving monthly anomalies of the ~~maximum streamflow deficit~~ evaporation (~~anomaly~~ET, ERA5-land). **(d)** Moving monthly anomalies of the soil moisture volume (ESM ERA5-land), soil moisture anomalies are depicted for a near-surface SM-level (black, 0–7 cm) and ~~event recovery~~the deepest level (blue, 100–289 cm) available from ERA5-Land. ~~Additional characteristics are~~ **(e)** Moving monthly anomalies of the snow water equivalent (SWE SPASS). **(f)** Moving monthly anomalies of the snowmelt (smlt, SPASS). **(g)** SPI colored by aggregation scales from 1- to 24-months. **(h)** SMRI colored by aggregation scales from 1- to 24-months. **(i)** Seven day average streamflow (M7Q) on which streamflow drought ~~intensity~~ events were identified. **(j)** The CQD time series shows the corresponding accumulated M7Q-deficits for both the fixed threshold approach (black) and the variable threshold approach (blue). **(k)** Absolute cumulative water deficit ~~volume or accumulated~~ (CWD). **(l)** Potential cumulative water deficit (PCWD). **(m)** Monthly anomalies of the CWD (CWD anomaly). **(n)** Monthly anomalies of the PCWD. Time series of the cumulative water deficits for both absolute values and ~~severity/magnitude~~ monthly anomalies are shown for both standard (~~maximum streamflow deficit~~black, P–E (P_e)) and snowmelt-corrected (blue, P–E+ΔSWE (PsmltSPASS_e)) variants. The ~~computation of other characteristics~~ same is ~~left to~~ shown for potential cumulative water deficits which are based on the ~~user~~potential water

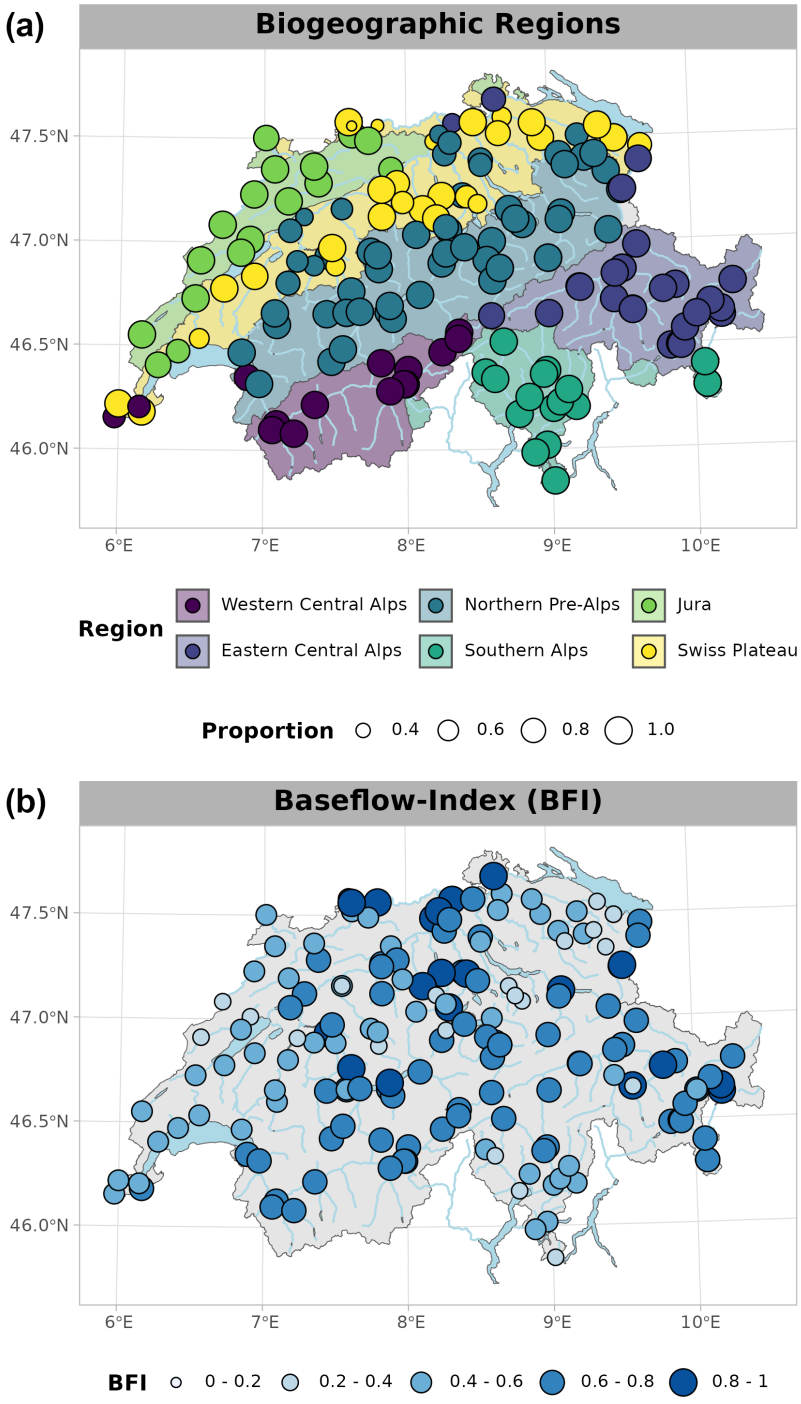

Figure 7. Catchment descriptors (examples). ~~**Top:**~~ (a) Dominant (largest overlap percentage with the catchment area) biogeographic region (colours). Point sizes indicate the catchment area proportion covered by the dominant biogeographic region. ~~**Bottom:**~~ (b) Baseflow-Index (BFI, Nathan and McMahon, 1990) for each catchment derived from the daily streamflow time series.

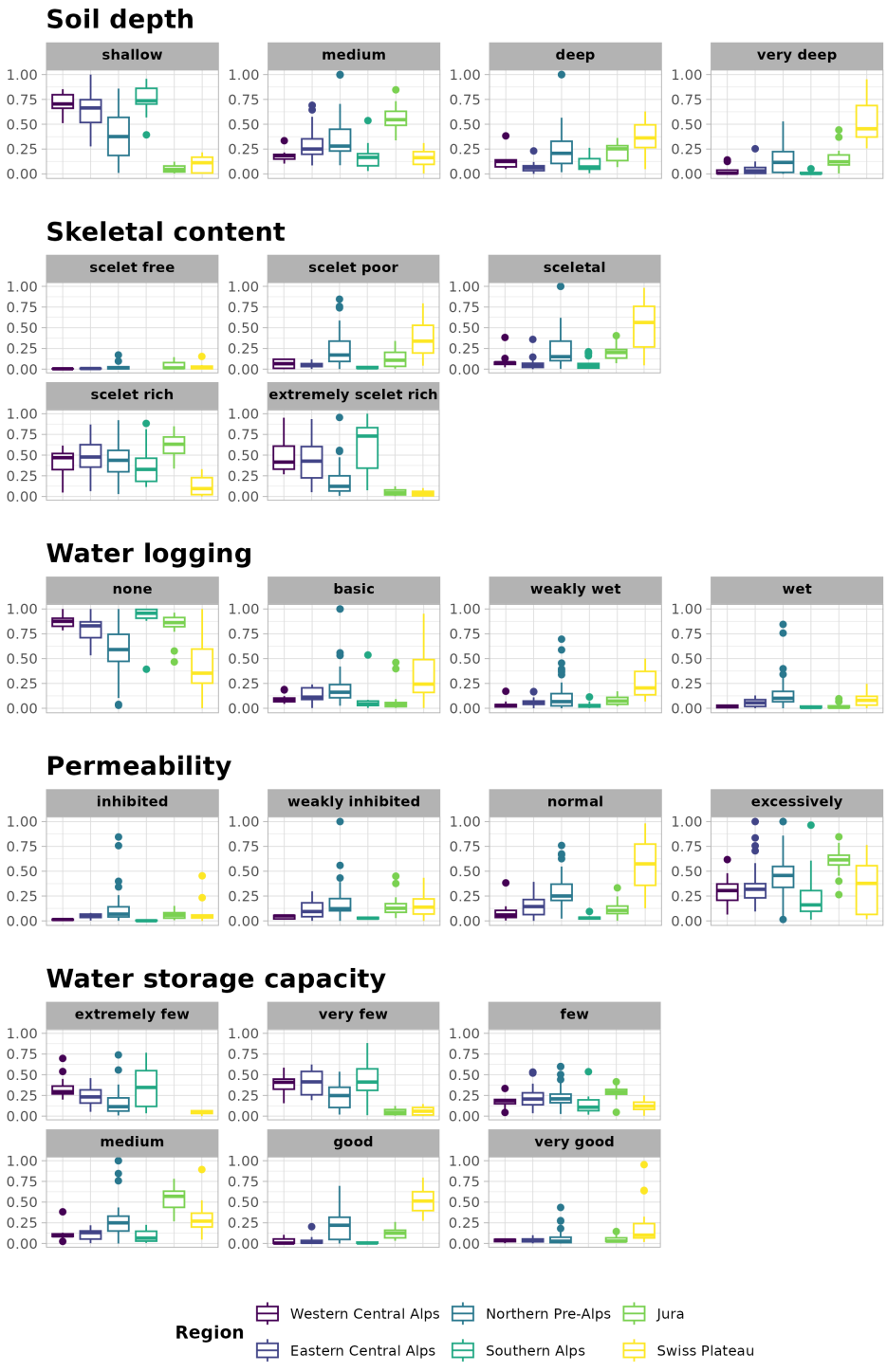

Figure A1. Catchment coverage fractions for all (sub-)categories of the soil characteristics: soil depth, skeletal content, water logging, soil permeability, and water storage capacity across regionalized catchment groups derived from the biogeographic regions of Switzerland (colours, see also Fig. 7a).

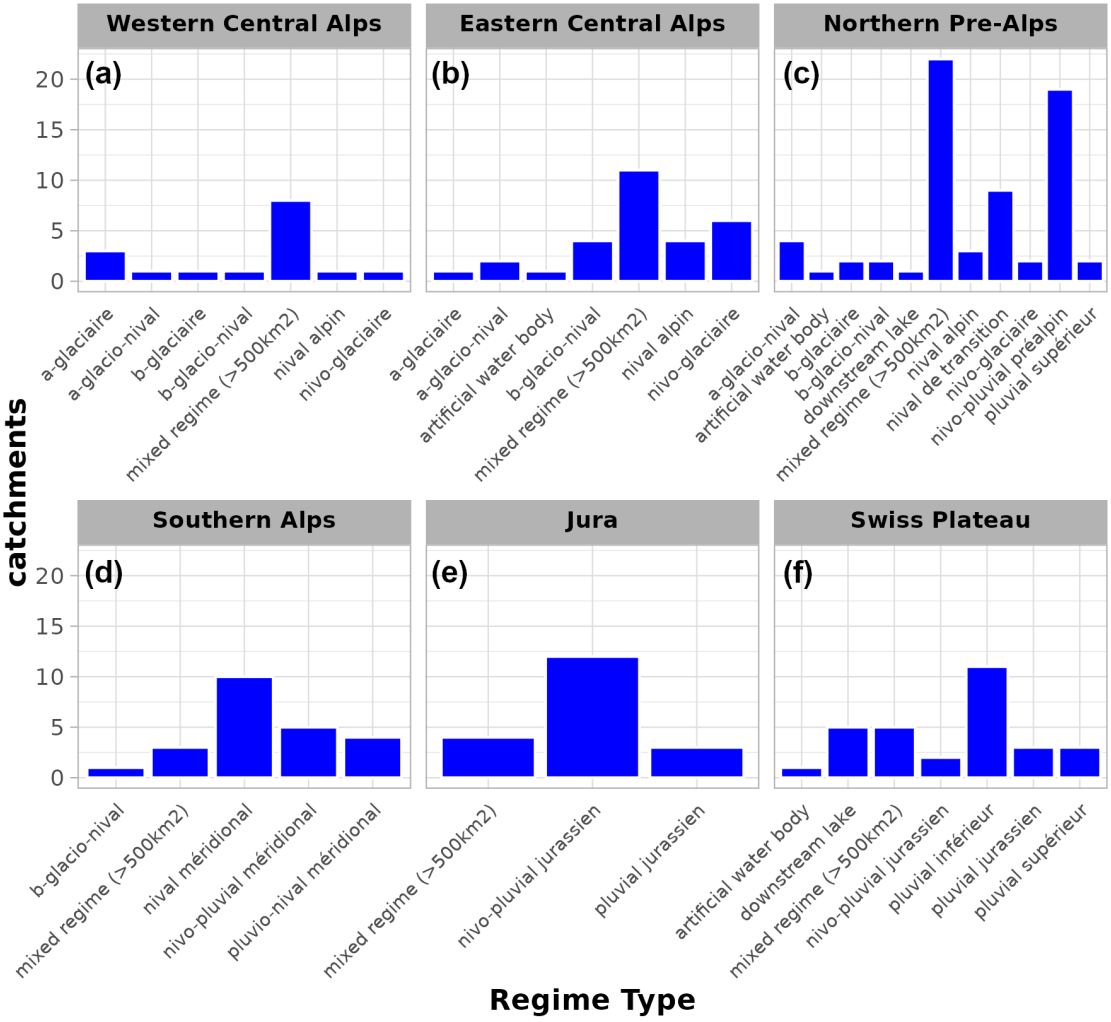

Figure A2. Streamflow regime type incidence among catchments grouped by the biogeographic regions of Switzerland (Western Central Alps, Eastern Central Alps, Northern Pre-Alps, Southern Alps, Jura and Swiss Plateau region; see Sect. 3.3 and also Fig. 7). The streamflow regime type classification was provided by the FOEN. The Y-axis shows the number of catchments associated with each category.

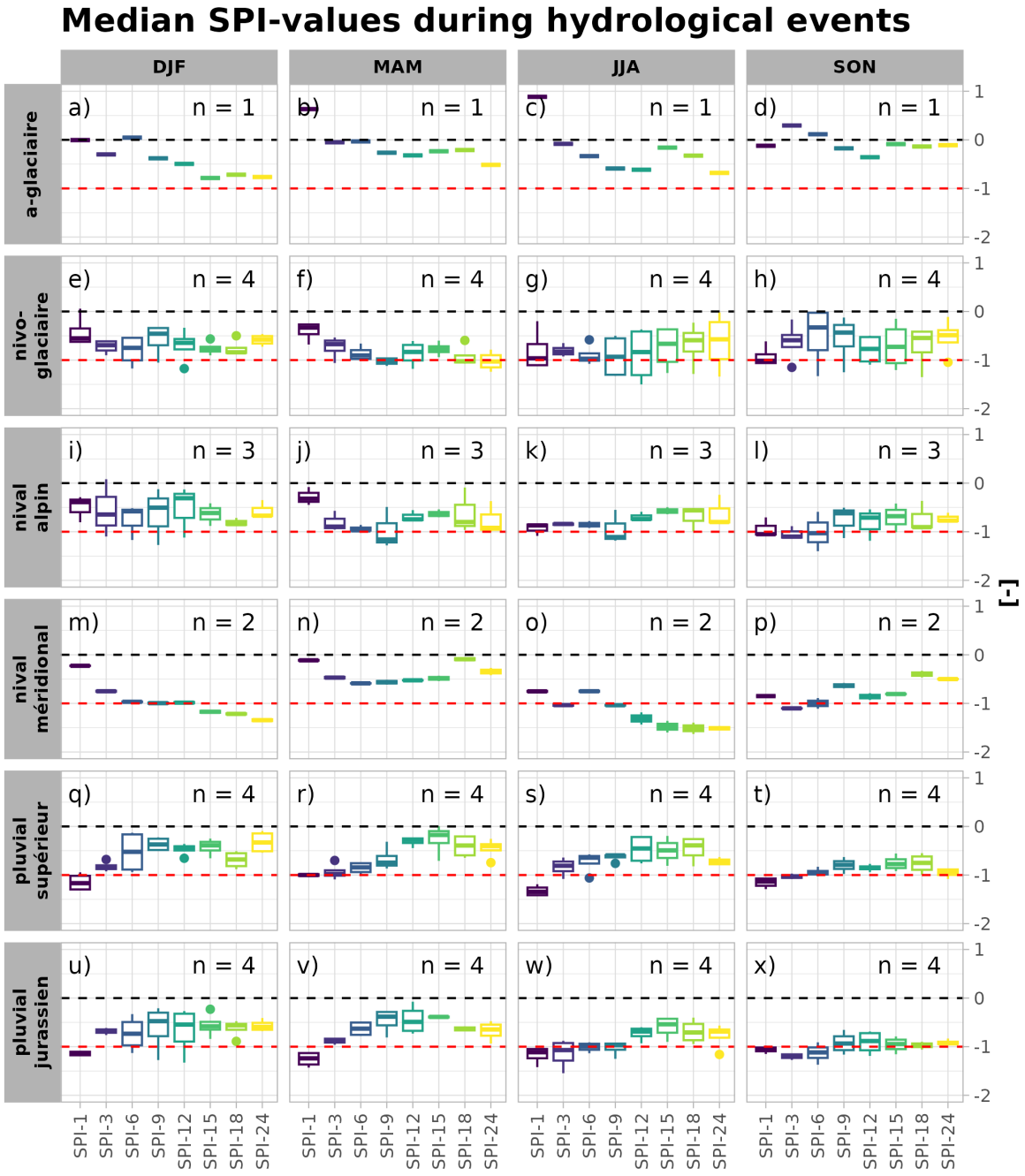

Figure A3. Median SPI values during hydrological drought conditions for all events of all catchments for six selected streamflow regime types across the four seasons winter (DJF), spring (MAM), summer (JJA) and autumn (SON). The streamflow regime types were selected to represent catchments with (dominant) glacial (a-glaciaire, nivo-glaciaire), snow (nival alpin, nival méridional) and pluvial processes (pluvial jurassien, pluvial supérieur) and spatial diversity. Hydrological drought events were defined by a moving monthly (31d) 15th-percentile (variable) threshold. Boxplots are coloured according to SPI aggregation time scales (1- to 24-months). Moderate drought conditions are indicated by the red dashed lines, the black dashed line indicates 0. "n=" refers to the number of catchments with a specific streamflow regime type.

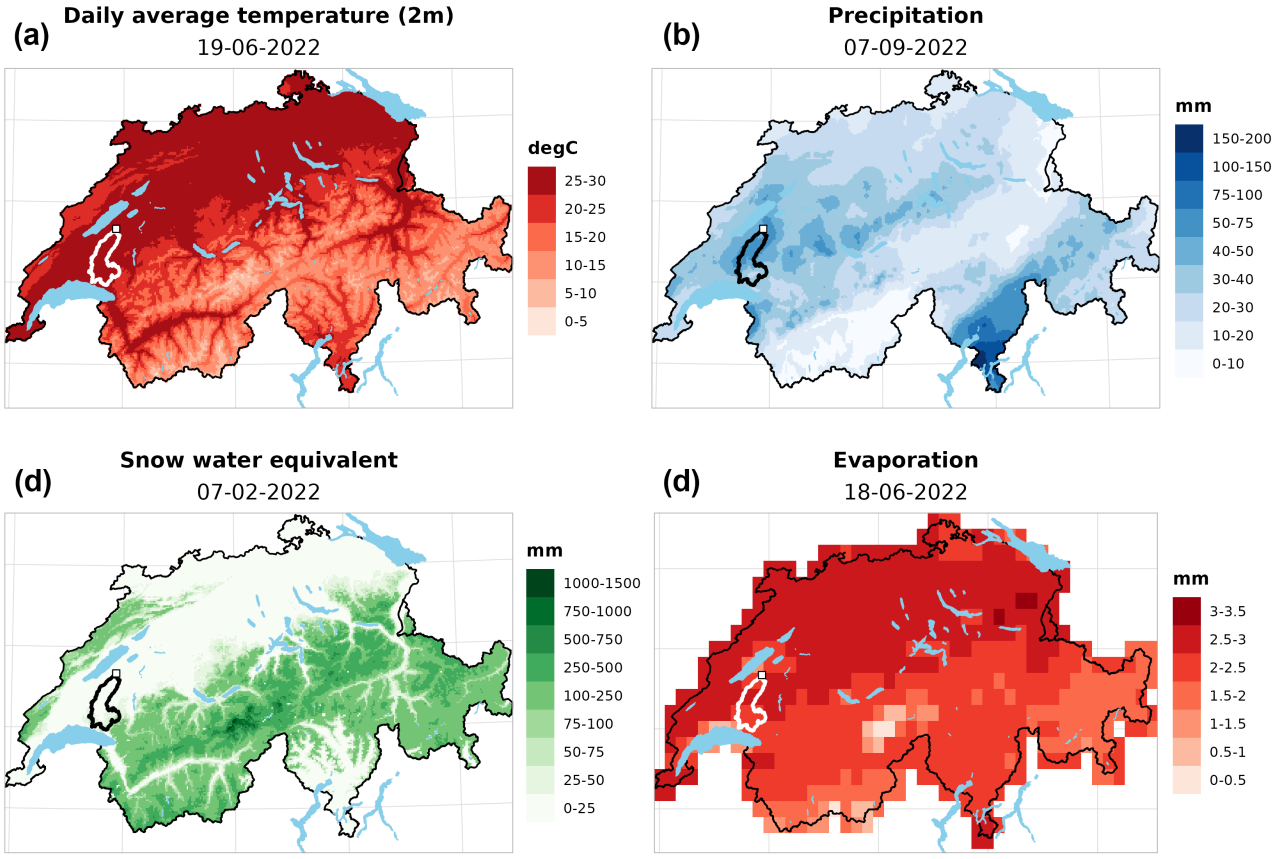

Figure A4. Overview of the spatial raster products used to extract daily time series. **(a)** Mean daily temperature (TabsD, MeteoSwiss), **(b)** Daily precipitation sun (RhiresD, MeteoSwiss), **(c)** Daily snow water equivalent of the Swiss snow climatology (SPASS) (SWE, MeteoSwiss & SLF), **(d)** Daily evaporation sum (aggregated from hourly ERA5-Land data, ECMWF). Note that the second snow climatology product (OSHD) is not shown. Contours in white/black show catchment *2034 - Broye, Payerne, Casernde d'aviation* for the day with the highest observed catchment average values for each specific product for the year 2022. White squares show the catchment outlet where daily streamflow is measured. Extracted and derived time series over the year 2022 are shown for the same catchment in Figure 6.

Appendix B: Tables

Appendix C: ~~Discussion~~

~~The HYD-RESPONSES dataset can, for example, be used to study drought dynamics and drought propagation, for streamflow forecasting using Long Short-Term Models Kratzert et al. (LSTMs; 2018); Lees et al. (LSTMs; 2022); Kratzert et al. (LSTMs; 2023), Random Forests Floriancic et al. (RFs; 2022), or to infer drought drivers using clustering and principal component analysis (e.g., Jehn et al., 2020). The information on breakpoints allows to study pre- and post-influence catchment behaviour e.g., by using the paired catchment or upstream-downstream approach (see e.g., Rangecroft et al., 2019; Van Loon et al., 2019) . The availability of variables originating from multiple data sources (direct observations, reanalysis data, model data) allows for comparative analyses. The following variables are available from multiple sources: temperature (MeteoSwiss, ERA5-Land), precipitation (MeteoSwiss, ERA5-Land), and snow (MeteoSwiss, WSL, ERA5-Land).There are several known limitations related to the datasets used to compile the HYD-RESPONSES data. ERA5-Land is a state-of-the-art reanalysis product provided at a higher spatial resolution than the standard ERA5 reanalysis (Hersbach et al., 2020; Muñoz-Sabater et al., 2021). The higher spatial resolution results in a better depiction of soil moisture, lakes, river discharge estimations, and the orographic enhancement of precipitation (Muñoz-Sabater et al., 2021) . However, the grid resolution of 9 km still has limitations over complex high-altitude terrain. The extracted time series related to snow depth (SWE) should be used with caution, as snow depth in ERA5-Land is of mixed quality depending on geographical location and altitude (Dalla Torre et al., 2024). Scherrer et al. (2023) showed, that ERA5-Land overestimates SWE at high elevations with larger biases in the southern compared to the northern Alps. They state that higher-resolution datasets such as SPASS (Marty et al., 2025) and OSHD (Mott, 2023; Mott et al., 2023) should be preferred over ERA5-Land. Further also note that all snow-related datasets have problems in representing small SWE amounts at low altitudes (Scherrer et al., 2023; Michel et al., 2023; Marty et al., 2025).Another limitation of the ERA5-Land dataset is the parameterization of subgrid-scale processes and the representation of subsurface storages that affect evapotranspiration (e.g., fixed maximum storage volume assumption; see Muñoz-Sabater et al., 2021). However, gridded observation-based evaporation datasets are yet to be developed for Switzerland. Caution is required when using the snow-corrected precipitation (water input) time series. The time series corrected by the ΔSWE series consider both snowfall ($\Delta SWE > 0$) and snowmelt ($\Delta SWE < 0$), while the correction based on snowmelt variables only accounts for snowmelt (*smlt* in ERA5-Land and *rome* in OSHD; see Table 1 and Table B1). Snowmelt-corrected precipitation time series only account for snowmelt, they may, therefore, be of limited use during the main snow accumulation season but can still provide valuable information during the snowmelt season.The time series for standardized (drought) indices are only provided for the transformation based on the best-fitting distribution across all catchments to allow for catchment comparability (see e.g., Staudinger et al., 2014) . The best-fitting distribution may however vary across catchments and climates (see e.g., Stagge et al., 2015). The HYD-RESPONSES dataset provides information on fits, missing values and flags which can be used to exclude catchments with unsatisfying fitting and transformation properties from analyses.The HYD-RESPONSES time series are provided for~~

product-specific periods and the spatial coverage is restricted to Swiss territory for most of the higher resolution MeteoSwiss and SLF products (TabsD, TminD, TmaxD, SPASS, SrelD, OSHD) as well as many catchment descriptor input datasets. Full coverage over the entire hydrological Switzerland is only available for ERA5-Land (all variables) and the MeteoSwiss RhiresD product (after 1992; see MeteoSwiss, 2021a). Catchments with significant areas outside of the Swiss National borders may therefore be considered with caution or excluded from the analysis.

Appendix C: Complementary datasets

Complementary datasets provide a wide range of additional catchment descriptors and hydro-meteorological time series. An overview of datasets and variables is provided in Table 4. The FOEN provides additional geodata related to both surface and groundwater via the Hydrological Service (). The datasets include additional catchment descriptors with information on population density, catchment areas covered by forest and agriculture (among others) as well as information on water quality aspects and sewage. The FOEN further operates both a groundwater monitoring network (NAQUA) providing continuous groundwater measurements for selected point locations (BAFU, 2019) and a water quality measurement network (NAWA) providing information on concentration and loads of important dissolved compounds (e.g., pH, electric conductivity, nutrient contents; BAFU, 2023).The "Catchment Attributes and Meteorology for Large-sample catchment Studies" (CAMELS) datasets aim at providing a consistent set of hydro-meteorological time series and catchment descriptors over a large sample of hydrological catchments on country level (Clerc-Schwarzenbach et al., 2024). The catchments in the Swiss version of the CAMELS data (CAMELS-CH; Höge et al., 2023a) are largely congruent with our dataset. The only exception is station 2646, which is only contained in the HYD-RESPONSES dataset. Note that the HYD-RESPONSES dataset provides only a sample subset of 184 catchments. The CAMELS-CH dataset provides valuable complementary catchment-level information on glacier changes (based on GLAMOS, for details see Höge et al., 2023a), land use, hydro-geological and hydro-terrestrial information (e.g., the contributions of various grain size categories and bulk-density) as well as anthropogenic disturbances (e.g., hydropower and reservoir capacities). CAMELS-CH further provides modelled time series based on the hydrological model PREVAH (see e.g., Höge et al., 2023a; Viviroli et al., 2009) . The CAMELS-CH dataset is freely available from Zenodo (https://zenodo.org/records/10354485; Höge et al., 2023b).The CombiPrecip dataset (MeteoSwiss) provides high-resolution (10 minutes, 1×1 km) precipitation fields derived from a combination of radar and station measurement data (Sideris et al., 2014). The CombiPrecip dataset could be a valuable addition for studying drought recovery where extreme precipitation is often considered an important factor (Wu et al., 2022).The HydCHeck project (Streeb et al., 2024) evaluated the influence of (anthropogenic) disturbance factors on streamflow at stations of the National Surface Water Quality (NAWA) Programme (BAFU, 2023). The evaluated NAWA stations are largely (87.5% of the stations) congruent with the HYD-RESPONSES dataset. The HydCHeck dataset provides catchment-level information on the magnitudes for all evaluated disturbance categories including water storage and regulation, hydropower, sewage water, constructions, agriculture as well as drinking and groundwater. The overall impact on several hydrological properties including low-, mid- and high-flow regimes as well as short-term effects and hydraulics is provided as categorical information (from

"not disturbed" to "strongly disturbed"). For more information see Streeb et al. (2024).As part of the planned Swiss National drought early warning system (DEWS), both a high-resolution remote-sensing based evaporation product (V. Humphrey, pers. comm.) and an automatic soil moisture measurement network are under development at MeteoSwiss, ETH Zurich and WSL and may become a valuable addition in a future.

Datasets compatible and complementary to the HYD-RESPONSES dataset. **DatasetShort descriptionProvider**Includes catchment proportions of forests, agricultural (crop) land, population, built-up area and more Groundwater measurement network (NAQUA) Groundwater measurements FOENWater quality measurement network (NAWA) Information on water quality parameters FOENSwiss version of the Catchment Attributes and Meteorology for Large sample catchment Studies (CAMELS) datasetHigh-resolution precipitation fields at ground based on a combination of radar and measurement data Detailed evaluation of influences and disturbances of the streamflow at NAWA measurement stations

The HYD-RESPONSES dataset contains data for 184 Swiss catchments that cover a variety of streamflow regimes, mean altitudes, catchment areas, and anthropogenic influences/disturbances. The catchments cover all biogeographic regions of Switzerland. The HYD-RESPONSES dataset provides daily streamflow data and daily hydro-meteorological time series extracted from gridded data products of MeteoSwiss (TabsD, RhiresD, TmaxD, TminD, SrelD), Meteoswiss and SLF (SPASS), SLF (OSHD) and ECMWF (ERA5-Land). The variables include temperature, precipitation, evaporation, sunshine duration, solar radiation, snowmelt, snow water equivalent, soil moisture, surface runoff, runoff, and streamflow. HYD-RESPONSES further provides derived variables related to streamflow (e.g., M7Q), water balance (e.g., P–E) and snowfall. Additionally, three standardized drought indices (SPI, SPEI, SMRI) for accumulation periods from 1 to 24 months and information on the (non-standardized) cumulative water deficit (CWD), the potential cumulative water deficit (PCWD) and cumulative streamflow deficit (CQD) are provided.The data set alsop provides information on (streamflow) drought events (occurrence and duration). For each catchment, the drought events have been identified based on fixed and on seasonally varying percentile thresholds.The combination of data sources, the information on hydro-meteorological variables (mainly temperature, precipitation and snow), the derived indices (water balance, cumulative water deficits, standardized drought indices, climatology and anomalies) allow for a multi-purpose use and various analytical approaches such as time series analysis (e.g., Kratzert et al., 2018; Lees et al., 2022), drought propagation and catchment sensitivity analysis (e.g., based on principal component analysis and clustering; Jehn et al., 2020) and changes in rainfall-runoff relationships during hydrological droughts (e.g., Wu et al., 2021). The HYD-RESPONSES data set can easily be combined with complementary datasets such as CAMELS-CH (Höge et al., 2023a) and HydCHeck (Streeb et al., 2024). The catchment time series vary in length (subject to station initialization), the hydrological time series are provided for the entire measurement period along with information on data homogeneity (see BAFU (2024) for more details).

Limitations exist for catchments extending beyond the Swiss borders. The catchment descriptors were extracted from datasets only covering Swiss national territory. The MeteoSwiss-based datasets cover only Switzerland except for RhiresD, which covers the entire hydrological Switzerland from 1992 onward. In summary, the data set provides a state-of-the-art data basis to study droughts in Switzerland.

The HYD-RESPONSES project was funded by the Federal Office for the Environment (FOEN). The preliminary SPASS dataset was kindly provided by Regula Muelchi (MeteoSwiss). We thank Caroline Kan (FOEN) for her help with the catchment selection.

Appendix C

Streamflow regime type incidence among catchments grouped by the biogeographic regions of Switzerland (Western Central Alps, Eastern Central Alps, Northern Pre-Alps, Southern Alps, Jura and Swiss Plateau region; see Section 3.3 and also Fig. 7). The streamflow regime type classification was provided by the FOEN.

Table B1. Glossary of extracted time series variables, their description and units.

Dataset	Variables	Variables (fullname)	Units	Producer
Spatial Climate Analyses	TabsD	Daily 2 m mean temperature	°C	MeteoSwiss
	RhiresD	Daily precipitation sums	mm	
	TminD	Daily 2 m minimum temperature	°C	
	TmaxD	Daily 2 m maximum temperature	°C	
	SrelD	Daily sunshine duration	%	
Snow Climatology for Switzerland (SPASS)	SWECLQMD	Daily snow water equivalent	mm	MeteoSwiss & SLF
Climatological snow data since 1998 (OSHD)	swee	Daily snow water equivalent	mm	SLF
	romc	Daily snowmelt-contribution to runoff	mm	
ERA5-Land	tp	Total precipitation	mm	ECMWF
	t2m	Average 2 m temperature	°C	
	e	Total evaporation	mm	
	pev	Total potential evaporation	mm	
	smlt	Snowmelt	mm	
	sd	Snow water equivalent	mm	
	ssr	Total solar radiation	MJ/m^2	
	ro	Runoff	mm	
	sro	Surface runoff	mm	
	swvl1	Soil water volume level 1 (0–7cm)	mm	
	swvl2	Soil water volume level 2 (7–28cm)	mm	
	swvl3	Soil water volume level 3 (28–100cm)	mm	
	swvl4	Soil water volume level 4 (100–289cm)	mm	
Streamflow time series	Q	Daily mean streamflow	m^3/s	FOEN

Table B2. Information on the underlying processing and reliability of basic and derived time series variables.

	Type	Level	Data source	Processing	Temporal resolution	Modeling component	Variables
Sect. 4.1	Streamflow	L1	Streamflow measurements (Q-Meas.)	–	daily mean	–	Q
	Precipitation	L2	Spatial climate analyses (SCA)	catchment average	daily total (sum)	–	RhiresD
	Sunshine Duration	L2		catchment average	daily relative (%)	–	SrelD
	Temperature	L2		catchment average	daily mean	–	TabsD, TmaxD, TminD
	Snow	L2	SPASS	catchment average	daily mean	yes	SWECLQMD
	Snow	L2	OSHD	catchment average	daily mean	yes	swee
		L2		catchment average	daily total (sum)	yes	romc
	Various	L2	ERA5-Land	catchment average, daily mean	daily mean	–	t2m
		L2		catchment average, daily mean	daily mean	strong	pev, sd, ssr, swvl1, swvl2, swvl3, swvl4
		L2		catchment average, daily total (sum)	daily total (sum)	–	tp
		L2		catchment average, daily total (sum)	daily total (sum)	strong	ro, smlt, sro, e
Sect. 4.2	Streamflow	L1	Q-Meas.	7-day average (centered)	daily	–	M7Q
		L3	ERA5-Land			strong	ro7Q
	Snow-related	L2	SCA, OSHD	P + Snowmelt	daily	yes	P_SMLT_OSHDromc
		L2	SCA, OSHD	P + ΔSWE	daily	yes	P_SMLT_OSHDswe
		L2	SCA, SPASS	P + ΔSWE	daily	yes	P_SMLT_SPASS
		L2	OSHD	ΔSWE	daily	yes	SWE_diff_OSHD
		L2	SPASS	ΔSWE	daily	yes	SWE_diff_SPASS
		L2	OSHD	ΔSWE > 0	daily	yes	SWE_posdiff_OSHD
		L2	SPASS	ΔSWE > 0	daily	yes	SWE_posdiff_SPASS
		L3	ERA5-Land	P + Snowmelt	daily	strong	tp_smlt
	Water balance	L3	SCA, ERA5-Land	P - E	daily	strong	P_e
		L3	SCA, ERA5-Land	P - PET	daily	strong	P_pev
		L3	SCA, OSHD, ERA5-Land	P + Snowmelt - E	daily	strong	PsmltOSHDromc_e
		L3	SCA, OSHD, ERA5-Land	P + Snowmelt - PET	daily	strong	PsmltOSHDromc_pev
		L3	SCA, OSHD, ERA5-Land	P + ΔSWE - E	daily	strong	PsmltOSHDswe_e
		L3	SCA, OSHD, ERA5-Land	P + ΔSWE - PET	daily	strong	PsmltOSHDswe_pev
		L3	SCA, SPASS, ERA5-Land	P + ΔSWE - E	daily	strong	PsmltSPASS_e
		L3	SCA, SPASS, ERA5-Land	P + ΔSWE - PET	daily	strong	PsmltSPASS_pev
		L3	ERA5-Land	P - E	daily	strong	tp_e
		L3	ERA5-Land	P - PET	daily	strong	tp_pev
		L3	ERA5-Land	P + Snowmelt - E	daily	strong	tpsmlt_e
		L3	ERA5-Land	P + Snowmelt - PET	daily	strong	tpsmlt_pev

Table B3. Information on the underlying processing and reliability of standardized (drought) indices and cumulative deficit time series.

	Type	Level	Data source	Processing	Temporal resolution	Distributions	Index Aggregation (months)	Modeling component	Variables
Sect. 4.4	SPI	L3	SCA	fitted transformation	daily	gamma	1 - 24	yes	RhiresD
		L3	ERA5-Land	fitted transformation	daily	gamma	1 - 24	yes	tp
	SPEI	L3	SCA, ERA5-Land	fitted transformation	daily	genlog	1 – 24	strong	P_pev
		L3	ERA5-Land	fitted transformation	daily	genlog	1 – 24	strong	tp_pev
	SMRI	L3	SCA, SPASS	fitted transformation	daily	genlog	1 – 24	yes	P_SMLT_SPASS
		L3	SCA, OSHD	fitted transformation	daily	lnorm	1 – 24	yes	P_SMLT_OSHDromc
		L3	SCA, OSHD	fitted transformation	daily	lnorm	1 – 24	yes	P_SMLT_OSHDswe
		L3	ERA5-Land	fitted transformation	daily	lnorm	1 – 24	strong	tp_smlt

	Type	Level	Data source	Processing	Temporal resolution	Variants	Threshold-level	Modeling component	Variables
Sect. 4.3	CWD	L3	SCA, ERA5-Land	cumulative sum of negative threshold deviations	daily	multi-year,yearly	P - E < 0	strong	P_e
		L3	SCA, OSHD, ERA5-Land		daily	multi-year,yearly	P - E < 0	strong	PsmltOSHDromc_e
		L3	SCA, OSHD, ERA5-Land		daily	multi-year,yearly	P - E < 0	strong	PsmltOSHDswe_e
		L3	SCA, SPASS, ERA5-Land		daily	multi-year,yearly	P - E < 0	strong	PsmltSPASS_e
		L3	ERA5-Land		daily	multi-year,yearly	P - E < 0	strong	tp_e
		L3	ERA5-Land		daily	multi-year,yearly	P - E < 0	strong	tpsmlt_e
	PCWD	L3	SCA, ERA5-Land	cumulative sum of negative threshold deviations	daily	multi-year,yearly	P - PET < 0	strong	P_pev
		L3	SCA, OSHD, ERA5-Land		daily	multi-year,yearly	P - PET < 0	strong	PsmltOSHDromc_pev
		L3	SCA, OSHD, ERA5-Land		daily	multi-year,yearly	P - PET < 0	strong	PsmltOSHDswe_pev
		L3	SCA, SPASS, ERA5-Land		daily	multi-year,yearly	P - PET < 0	strong	PsmltSPASS_pev
		L3	ERA5-Land		daily	multi-year,yearly	P - PET < 0	strong	tp_pev
		L3	ERA5-Land		daily	multi-year,yearly	P - PET < 0	strong	tpsmlt_pev
Sect. 4.6	CQD	L2	Q-Meas.	cumulative sum of negative threshold deviations	daily	multi-year,yearly	**monthly/seasonal:** 2^{nd}, 5^{th}, 10^{th}, 15^{th}, 25^{th}, 50^{th} (median), mean **yearly:** Q347	–	M7Q

Table B4. Information on the underlying processing and reliability of climatology and anomaly time series.

	Type	Level	Processing	Temporal resolution	Variant	Statistics	Window / Anomaly-Scale	Variables
Sect. 4.5	climatology	L1–L2	statistical summary	monthly seasonal extended season yearly	regular	min, q05, q25, med, mean, q75, q95, max, sd, sum	–	Sect. 4.1
	climatology	L1–L3	statistical summary	monthly seasonal extended season yearly	regular	min, q05, q25, med, mean, q75, q95, max, sd, sum	–	Sect. 4.2
	climatology	L2–L3	statistical summary	monthly seasonal extended season yearly	regular	min, q05, q25, med, mean, q75, q95, max, sd, sum	–	Sect. 4.3
	climatology	L2	statistical summary	monthly seasonal extended season yearly	regular	min, q05, q25, med, mean, q75, q95, max, sd, sum	–	Sect. 4.6
	climatology	L1–L2	statistical summary	daily	window	min, q05, q25, med, mean, q75, q95, max, sd, sum	no window (daily) 31 days (monthly) 91 days (seasonal)	Sect. 4.1
	climatology	L1–L3	statistical summary	daily	window	min, q05, q25, med, mean, q75, q95, max, sd, sum	no window (daily) 31 days (monthly) 91 days (seasonal)	Sect. 4.2
	climatology	L2–L3	statistical summary	daily	window	min, q05, q25, med, mean, q75, q95, max, sd, sum	no window (daily) 31 days (monthly) 91 days (seasonal)	Sect. 4.3
	climatology	L2	statistical summary	daily	window	min, q05, q25, med, mean, q75, q95, max, sd, sum	no window (daily) 31 days (monthly) 91 days (seasonal)	Sect. 4.6
	anomalies	L1–L2	z-scores	daily	window	(value - mean) / sd	no window (daily) 31 days (monthly) 91 days (seasonal)	Sect. 4.1
	anomalies	L1–L3	z-scores	daily	window	(value - mean) / sd	no window (daily) 31 days (monthly) 91 days (seasonal)	Sect. 4.2
	anomalies	L2–L3	z-scores	daily	window	(value - mean) / sd	no window (daily) 31 days (monthly)	Sect. 4.3

Table B5. Information on the underlying processing and reliability of event time series.

	Type	Level	Data source	Processing	Temporal resolution	Threshold-level	Variables
Sect. 4.7	Cumulative water deficits	L3	Sect. 4.3	deficit > 0	daily	–	Sect. 4.3
	Cumulative streamflow deficits	L2	Sect. 4.6	deficit > 0	daily	–	Sect. 4.6
	Streamflow droughts	L2	FOEN	M7Q < threshold-level	daily	**monthly/seasonal:** 2^{nd}, 5^{th}, 10^{th}, 15^{th}, 25^{th}, 50^{th} (median), mean **yearly:** Q347	Sect. 4.7

Table B6. Characteristics of all 184 catchments in the HYD-RESPONSES dataset (Part 1/5).

Catchment	Water name	Place	Lon / Lat	Glaciation	Area	Avg. Height	Regime Type	Yearly Avg. T	Yearly P	Yearly E	Yearly Q
			EPSG:21781	%	km^2	m asl		°C	mm	mm	mm
0070	Emme	Emmenmatt	623610 / 200420	0.0	443.00	1065	nivo-pluvial préalpin	6.94	1539.26	300.29	856.45
0078	Poschiavino	Le Prese	803490 / 130520	4.0	168.00	2162	nival méridional	1.58	1324.78	175.84	1078.26
0155	Emme	Wiler, Limpachmündung	608220 / 223240	0.0	937.00	858	mixed regime (>500km^2)	7.80	1356.58	313.03	623.66
0185	Plessur	Chur	757975 / 191925	0.0	264.00	1868	nival alpin	3.42	1179.00	194.41	915.28
0308	Goldach	Goldach, Bleiche	753190 / 261590	0.0	51.10	827	pluvial supérieur	8.33	1423.14	317.59	889.48
0352	Linth	Linthal, Ausgleichsbecken KLL	718285 / 197310	9.4	147.00	2085	a-glacio-nival	1.83	1874.45	169.98	2422.14
0403	Inn	Cinuos-chel	797700 / 168170	5.2	733.00	2456	mixed regime (>500km^2)	-0.66	1007.10	146.31	1007.68
0488	Simme	Latterbach	610680 / 167840	1.5	563.00	1594	nival de transition	4.79	1506.56	225.96	1108.96
0491	Schächen	Bürglen, Galgenwäldli	692480 / 191800	1.5	108.00	1728	nivo-glaciaire	3.55	1854.00	188.20	1646.44
2009	Rhône	Porte du Scex	557660 / 133280	11.0	5238.00	2127	mixed regime (>500km^2)	1.96	1292.71	148.24	1116.86
2011	Rhône	Sion	593770 / 118630	14.2	3372.00	2291	mixed regime (>500km^2)	1.00	1240.59	127.84	966.98
2016	Aare	Brugg	657000 / 259360	1.5	11681.00	1000	mixed regime (>500km^2)	7.34	1317.32	291.55	833.95
2018	Reuss	Mellingen	662830 / 252580	1.8	3386.00	1259	mixed regime (>500km^2)	5.98	1592.79	239.54	1300.70
2019	Aare	Brienzwiler	649930 / 177380	15.5	555.00	2135	mixed regime (>500km^2)	1.41	1842.89	123.14	2077.41
2020	Ticino	Bellinzona	721245 / 117025	0.2	1517.00	1679	mixed regime (>500km^2)	4.27	1658.40	209.25	1339.14
2024	Rhône	Branson	573150 / 108300	13.0	3728.00	2235	mixed regime (>500km^2)	1.35	1249.40	133.83	1166.41
2029	Aare	Brügg, Aegerten	588220 / 219020	2.1	8249.00	1142	mixed regime (>500km^2)	6.73	1366.01	276.35	908.00
2030	Aare	Thun	613230 / 179280	6.9	2459.00	1746	mixed regime (>500km^2)	3.72	1604.05	176.90	1438.59
2033	Vorderrhein	Ilanz	735000 / 182030	1.8	774.00	2030	mixed regime (>500km^2)	2.28	1534.35	157.08	1340.76
2034	Broye	Payerne, Caserne d'aviation	561660 / 187320	0.0	416.00	715	pluvial inférieur	9.19	1186.40	322.34	564.59
2044	Thur	Andelfingen	693510 / 272500	0.0	1702.00	770	mixed regime (>500km^2)	8.25	1392.81	333.74	857.45
2053	Drance	Martigny, Pont de Rossettan	570930 / 105200	11.3	676.00	2250	mixed regime (>500km^2)	1.40	1269.62	138.70	462.70
2056	Reuss	Seedorf	690085 / 193210	6.4	833.00	2013	mixed regime (>500km^2)	1.96	1681.87	150.85	1624.38
2063	Aare	Murgenthal	629530 / 235090	1.7	10059.00	1066	mixed regime (>500km^2)	7.04	1346.17	284.49	888.26
2070	Emme	Emmenmatt, nur Hauptstation	623610 / 200420	0.0	443.00	1065	nivo-pluvial préalpin	6.94	1539.26	300.29	835.36
2078	Poschiavino	Le Prese, stazione principale	803490 / 130520	4.0	168.00	2162	nival méridional	1.58	1324.78	175.84	1064.96
2084	Muota	Ingenbohl	688230 / 206140	0.0	317.00	1363	nival de transition	5.45	1958.67	237.00	1915.23
2085	Aare	Hagneck	580680 / 211650	3.4	5112.00	1368	mixed regime (>500km^2)	5.61	1452.15	237.22	1068.53
2086	Brenno	Loderio	717770 / 137270	0.3	400.00	1815	nival méridional	3.68	1618.84	185.01	342.59
2087	Reuss	Andermatt	688120 / 166320	2.9	190.00	2284	b-glacio-nival	0.59	1709.03	116.01	1167.49
2091	Rhein	Rheinfelden, Messstation	627190 / 267840	0.8	34524.00	1068	mixed regime (>500km^2)	6.68	1351.54	281.92	935.92
2099	Limmat	Zürich, Unterhard	682055 / 249430	0.8	2174.00	1194	mixed regime (>500km^2)	6.28	1719.03	264.98	1353.24
2102	Sarner Aa	Sarnen	661460 / 194220	0.0	269.00	1281	downstream lake	5.99	1648.90	224.99	1167.84
2104	Linth	Weesen, Biäsche	725160 / 221380	1.6	1062.00	1584	mixed regime (>500km^2)	4.39	1785.64	221.45	1538.41
2105	Inn	St. Moritzbad	783910 / 150960	3.8	155.00	2399	b-glacio-nival	-0.33	1055.10	161.39	1145.54
2106	Birs	Münchenstein, Hofmatt	613570 / 263080	0.0	887.00	728	mixed regime (>500km^2)	8.53	1206.82	335.73	545.50
2109	Lütschine	Gsteig	633130 / 168200	13.5	381.00	2050	a-glacio-nival	2.11	1780.73	119.63	1580.50
2110	Reuss	Mühlau, Hünenberg	672520 / 230600	2.2	2902.00	1371	mixed regime (>500km^2)	5.42	1641.78	226.17	1399.39
2112	Sitter	Appenzell	749040 / 244220	0.1	74.40	1256	nival de transition	6.22	1896.65	345.94	1421.58
2117	Drance de Bagnes	Le Châble, Villette	582550 / 103270	22.1	254.00	2609	b-glaciaire	-0.59	1274.15	118.82	254.59
2119	Sarine	Fribourg	579420 / 183670	0.2	1271.00	1247	mixed regime (>500km^2)	6.35	1420.49	276.29	975.93
2122	Birse	Moutier, La Charrue	595740 / 237010	0.0	186.00	921	nivo-pluvial jurassien	7.49	1371.76	333.62	520.67

T = Temperature, P = Precipitation, E = Evaporation, Q = Streamflow/Runoff

Table B7. Characteristics of all 184 catchments in the HYD-RESPONSES dataset (Part 2/5).

Catchment	Water name	Place	Lon / Lat	Glaciation	Area	Avg. Height	Regime Type	Yearly Avg. T	Yearly P	Yearly E	Yearly Q
			EPSG:21781	%	km^2	m asl		°C	mm	mm	mm
2125	Lorze	Frauenthal	674715 / 229845	0.0	262.00	678	downstream lake	8.91	1427.56	309.43	911.18
2126	Murg	Wängi	714105 / 261720	0.0	80.20	652	pluvial inférieur	8.72	1282.82	340.64	693.21
2132	Töss	Neftenbach	691460 / 263820	0.0	343.00	658	pluvial inférieur	8.89	1331.52	339.46	701.10
2135	Aare	Bern, Schönau	600710 / 198000	5.8	2941.00	1596	mixed regime (>500km^2)	4.43	1542.51	196.01	1317.74
2139	Rheintaler Binnenkanal	St. Margrethen	767160 / 257780	0.0	175.00	710	artificial waterbody	9.01	1451.16	310.08	2038.65
2141	Albula	Tiefencastel	763420 / 170145	0.5	529.00	2128	mixed regime (>500km^2)	1.53	1018.61	159.38	904.27
2143	Rhein	Rekingen	667060 / 269230	0.2	14767.00	1131	mixed regime (>500km^2)	5.83	1296.20	276.20	945.75
2150	Landquart	Felsenbach	765365 / 204910	0.7	614.00	1797	mixed regime (>500km^2)	3.34	1289.45	188.82	1203.19
2151	Simme	Oberwil	600060 / 167090	2.4	344.00	1641	nival de transition	4.52	1536.17	215.10	1084.54
2152	Reuss	Luzern, Geissmattbrücke	665330 / 211800	2.8	2254.00	1504	mixed regime (>500km^2)	4.72	1683.08	207.59	1526.87
2155	Emme	Wiler, Limpachmündung, nur Hauptstation	608220 / 223240	0.0	924.00	863	mixed regime (>500km^2)	7.80	1356.58	313.03	316.15
2159	Gürbe	Belp, Mülimatt	604810 / 192680	0.0	116.00	846	pluvial supérieur	8.05	1236.50	298.87	715.53
2160	Sarine	Broc, Château d'en bas	573520 / 161345	0.3	636.00	1500	mixed regime (>500km^2)	5.12	1500.63	248.29	1014.01
2161	Massa	Blatten bei Naters	643700 / 137290	56.5	196.00	2937	a-glaciaire	-2.89	2036.03	45.98	2433.50
2167	Tresa	Ponte Tresa, Rocchetta	709580 / 92145	0.0	609.00	803	mixed regime (>500km^2)	9.59	1789.28	351.80	1107.04
2170	Arve	Genève, Bout du Monde	501220 / 115120	5.1	1973.00	1370	mixed regime (>500km^2)	5.65	1505.42	249.36	1153.49
2174	Rhône	Chancy, Aux Ripes	486600 / 112340	6.6	10308.00	1569	mixed regime (>500km^2)	4.32	1324.91	216.70	1027.76
2176	Sihl	Zürich, Sihlhölzli	682145 / 246890	0.0	343.00	1045	nivo-pluvial préalpin	6.97	1787.58	294.48	621.39
2179	Sense	Thörishaus, Sensematt	593350 / 193020	0.0	351.00	1071	nivo-pluvial préalpin	7.22	1404.55	306.68	756.67
2181	Thur	Halden	733560 / 263180	0.0	1085.00	908	mixed regime (>500km^2)	7.61	1585.63	329.74	1082.92
2185	Plessur	Chur, nur Hauptstation	757975 / 191925	0.0	264.00	1868	nival alpin	3.42	1179.00	194.41	693.27
2187	Werdenberger Binnenkanal	Salez	756795 / 234005	0.0	183.00	1003	artificial waterbody	7.74	1547.48	279.64	1344.65
2199	Wiese	Basel	611800 / 269700	0.0	442.00	720	pluvial jurassien	10.67	1508.42	342.75	800.00
2200	Weisse Lütschine	Zweilütschinen	635310 / 164550	13.1	165.00	2165	a-glacio-nival	1.58	1767.24	112.23	1531.27
2202	Ergolz	Liestal	622270 / 259750	0.0	261.00	588	pluvial jurassien	9.55	1076.57	341.74	436.87
2203	Grande Eau	Aigle	563975 / 129825	0.8	132.00	1562	nival de transition	4.96	1617.15	240.72	1082.21
2205	Aare	Untersiggenthal, Stilli	659970 / 263180	1.4	17553.00	1064	mixed regime (>500km^2)	6.99	1416.13	279.19	984.66
2206	Melera	Melera (Valle Morobbia)	726988 / 114670	0.0	1.07	1423	nivo-pluvial méridional	5.88	1712.31	290.07	1297523.71
2210	Doubs	Ocourt	572530 / 244460	0.0	1275.00	952	mixed regime (>500km^2)	7.10	1499.13	346.76	790.06
2215	Saane	Laupen	584440 / 195300	0.1	1862.00	1137	mixed regime (>500km^2)	6.87	1373.00	288.00	866.85
2219	Simme	Oberried / Lenk	602630 / 141660	22.6	34.80	2347	b-glaciaire	0.92	1779.85	171.95	2130.36
2232	Allenbach	Adelboden	608710 / 148300	0.0	28.80	1863	nival alpin	3.64	1557.08	174.36	1332.39
2239	Spöl	Punt dal Gall	811020 / 167920	0.3	295.00	2389	nivo-glaciaire	-0.61	940.86	152.75	105.09
2243	Limmat	Baden, Limmatpromenade	665640 / 258690	0.7	2394.00	1131	mixed regime (>500km^2)	6.59	1662.82	272.41	1311.73
2244	Krummbach	Klusmatten	644500 / 119420	0.4	19.40	2271	nival méridional	1.35	1342.63	166.60	1232.18
2247	Doubs	Sortie du lac des Brenets	544560 / 214880	0.0	867.00	977	mixed regime (>500km^2)	6.41	1548.43	350.16	635.02
2251	Rotenbach	Plaffeien, Schwyberg	587980 / 170590	0.0	1.69	1455	nival de transition	5.70	1688.19	296.97	1427822.37
2252	Schwändlibach	Plaffeien, Schwyberg	588340 / 171015	0.0	1.38	1439	nival de transition	5.76	1662.14	296.97	832954.23
2256	Rosegbach	Pontresina	788810 / 151690	21.7	66.50	2704	a-glaciaire	-1.75	1137.41	119.24	1398.90
2262	Berninabach	Pontresina	789440 / 151320	14.4	107.00	2615	a-glacio-nival	-1.18	1203.87	131.81	1411.84
2263	Chamuerabach	La Punt-Chamues-ch	791430 / 160600	0.1	73.40	2548	nivo-glaciaire	-1.10	1011.37	145.82	930.92
2265	Inn	Tarasp	816800 / 185910	3.0	1581.00	2384	mixed regime (>500km^2)	-0.37	992.18	147.59	383.43
2268	Rhone	Gletsch	670810 / 157200	41.8	39.40	2710	a-glaciaire	-1.75	1937.62	75.90	2342.03
2269	Lonza	Blatten	629130 / 140910	24.7	77.40	2624	a-glaciaire	-1.28	1566.54	86.75	1924.12
2276	Grosstalbach	Isenthal	685500 / 196050	6.7	43.90	1819	nival alpin	3.28	1731.55	227.73	1270.99

T = Temperature, P = Precipitation, E = Evaporation, Q = Streamflow/Runoff

Table B8. Characteristics of all 184 catchments in the HYD-RESPONSES dataset (Part 3/5).

Catchment	Water name	Place	Lon / Lat EPSG:21781	Glaciation %	Area km^2	Avg. Height m asl	Regime Type	Yearly Avg. T °C	Yearly P mm	Yearly E mm	Yearly Q mm
2282	Sperbelgraben	Wasen, Kurzeneialp	630725 / 207270	0.0	0.56	1070	nivo-pluvial préalpin	7.06	1631.78	342.53	883404.51
2283	Rappengraben	Wasen, Riedbad	634340 / 207350	0.0	0.60	1142	nivo-pluvial préalpin	6.87	1656.81	355.40	1064734.56
2288	Rhein	Neuhausen, Flurlingerbrücke	689145 / 281975	0.3	11930.00	1239	mixed regime (>500km^2)	4.59	1295.50	261.93	960.66
2289	Rhein	Basel, Rheinhalle	613400 / 267650	0.8	35878.00	1052	mixed regime (>500km^2)	6.78	1343.83	284.04	919.47
2290	Areuse	St-Sulpice	532980 / 195880	0.0	104.00	1110	nivo-pluvial jurassien	5.67	1500.18	344.81	1408.21
2299	Alpbach	Erstfeld, Bodenberg	688560 / 185120	19.7	20.70	2205	b-glaciaire	1.07	1669.66	171.48	2406.18
2300	Minster	Euthal, Rüti	704425 / 215310	0.0	59.10	1352	nival de transition	5.53	2115.46	259.83	1639.61
2303	Thur	Jonschwil, Mühlau	723675 / 252720	0.0	493.00	1021	nivo-pluvial préalpin	6.93	1757.19	320.59	1285.82
2304	Ova dal Fuorn	Zernez, Punt la Drossa	810560 / 170790	0.0	55.30	2327	nival alpin	-0.46	937.69	150.89	586.32
2305	Glatt	Herisau, Zellersmühle	737270 / 251290	0.0	16.70	829	pluvial supérieur	8.18	1491.95	329.49	1063.60
2307	Suze	Sonceboz	579810 / 227350	0.0	127.00	1036	nivo-pluvial jurassien	6.97	1332.88	340.21	1008.10
2308	Goldach	Goldach, Bleiche, nur Hauptstation	753190 / 261590	0.0	50.40	832	pluvial supérieur	8.33	1423.14	317.59	853.11
2312	Aach	Salmsach, Hungerbühl	744410 / 268400	0.0	47.40	467	pluvial inférieur	9.68	1019.41	335.09	488.00
2319	Ova da Cluozza	Zernez	804930 / 174830	0.0	27.00	2371	nivo-glaciaire	-0.47	919.61	150.80	888.70
2321	Cassarate	Pregassona	718010 / 97380	0.0	75.80	987	pluvio-nival méridional	8.52	1900.06	330.75	983.33
2327	Dischmabach	Davos, Kriegsmatte	786220 / 183370	0.7	42.90	2376	b-glacio-nival	0.15	1015.77	147.17	1242.21
2342	Saltina	Brig	642220 / 129630	2.5	76.50	2014	nivo-glaciaire	2.53	1165.38	167.60	948.67
2343	Langete	Huttwil, Häberenbad	629560 / 219135	0.0	59.90	760	pluvial inférieur	8.14	1276.02	329.38	618.66
2346	Rhone	Brig	641340 / 129700	19.2	906.00	2339	mixed regime (>500km^2)	0.31	1630.34	103.68	1481.12
2347	Riale di Roggiasca	Roveredo, Bacino di compenso	733545 / 118160	0.0	8.12	1702	nivo-pluvial méridional	4.11	1684.56	288.12	1869.20
2349	Breggia	Chiasso, Ponte di Polenta	722315 / 78320	0.0	47.10	933	pluvio-nival méridional	8.58	1726.83	382.64	712.61
2351	Vispa	Visp	634050 / 125900	23.1	786.00	2648	mixed regime (>500km^2)	-0.92	1125.42	116.03	684.02
2352	Linth	Linthal, Ausgleichsbecken KLL, nur Haupt	718285 / 197310	9.4	147.00	2085	a-glacio-nival	1.83	1874.45	169.98	925.43
2355	Landwasser	Davos, Frauenkirch	779640 / 181200	0.3	184.00	2224	nivo-glaciaire	0.97	1063.29	151.03	926.33
2356	Riale di Calneggia	Cavergno, Pontit	684970 / 135960	0.0	23.90	2003	nival méridional	3.08	1868.56	188.46	1922.63
2364	Ticino	Piotta	694610 / 152450	0.3	159.00	2071	nival méridional	2.16	1803.44	136.09	413.29
2366	Poschiavino	La Rösa	802120 / 142010	0.0	14.10	2285	nival méridional	0.86	1398.05	162.22	1189.26
2368	Maggia	Locarno, Solduno	703100 / 113860	0.3	927.00	1530	mixed regime (>500km^2)	5.63	1946.42	245.77	783.83
2369	Mentue	Yvonand, La Mauguettaz	545440 / 180875	0.0	105.00	675	pluvial jurassien	9.35	1081.13	328.32	457.37
2370	Doubs	Le Noirmont, La Goule	561430 / 231050	0.0	1047.00	977	mixed regime (>500km^2)	6.69	1534.75	348.30	795.29
2371	Orbe	Le Chenit, Frontière	501445 / 156305	0.0	45.90	1235	nivo-pluvial jurassien	6.42	1901.34	337.96	615.79
2372	Linth	Mollis, Linthbrücke	723985 / 217965	2.9	600.00	1743	mixed regime (>500km^2)	3.49	1848.95	197.67	1687.89
2374	Necker	Mogelsberg, Aachsäge	727110 / 247290	0.0	88.10	956	nivo-pluvial préalpin	7.27	1718.29	338.15	1142.38
2378	Orbe	Orbe, Le Chalet	530080 / 175560	0.0	343.00	1139	nivo-pluvial jurassien	6.73	1692.73	341.81	1016.62
2386	Murg	Frauenfeld	709540 / 269660	0.0	213.00	597	pluvial inférieur	8.98	1178.35	343.21	567.47
2387	Hinterrhein	Fürstenau	753570 / 175730	0.6	1577.00	2127	mixed regime (>500km^2)	1.47	1147.93	165.74	767.37
2403	Inn	Cinuos-chel, nur Hauptstation	797700 / 168170	5.2	733.00	2456	mixed regime (>500km^2)	-0.66	1007.10	146.31	212.71
2409	Emme	Eggiwil, Heidbüel	627910 / 191180	0.0	124.00	1281	nivo-pluvial préalpin	6.10	1604.24	270.21	1061.40
2410	Liechtensteiner Binnenkanal	Ruggell	757750 / 234590	0.0	116.00	853	artificial waterbody	8.58	1286.29	264.96	1321.13
2412	Sionge	Vuippens, Château	572420 / 167540	0.0	43.40	865	nivo-pluvial préalpin	8.10	1298.00	302.73	802.84
2414	Rietholzbach	Mosnang, Rietholz	718840 / 248440	0.0	3.19	794	pluvial supérieur	8.13	1476.78	336.69	1006909.39
2415	Glatt	Rheinsfelden	678040 / 269720	0.0	417.00	503	downstream lake	9.70	1165.36	340.63	590.10
2416	Aabach	Hitzkirch, Richensee	661390 / 230220	0.0	73.30	581	downstream lake	9.46	1163.00	317.37	535.90
2417	Suhre	Oberkirch	651320 / 223140	0.0	75.60	583	downstream lake	9.39	1139.68	312.72	510.78
2418	Julia	Tiefencastel	763570 / 169910	0.2	325.00	2196	nivo-glaciaire	1.10	1058.77	161.39	97.09
2419	Rhone	Reckingen	661910 / 146780	11.8	214.00	2305	a-glacio-nival	0.30	1814.00	105.51	1424.61
2420	Moesa	Lumino, Sassello	724765 / 120360	0.1	472.00	1667	nivo-pluvial méridional	3.98	1619.63	234.73	1339.42

T = Temperature, P = Precipitation, E = Evaporation, Q = Streamflow/Runoff

Table B9. Characteristics of all 184 catchments in the HYD-RESPONSES dataset (Part 4/5).

Catchment	Water name	Place	Lon / Lat	Glaciation	Area	Avg. Height	Regime Type	Yearly Avg. T	Yearly P	Yearly E	Yearly Q
			EPSG:21781	%	km^2	m asl		°C	mm	mm	mm
2426	Seez	Mels	750410 / 212510	0.1	106.00	1803	nival alpin	3.55	1578.39	217.42	640.88
2430	Rein da Sumvitg	Sumvitg, Encardens	718810 / 167690	1.7	21.80	2457	b-glacio-nival	-0.15	1581.92	160.82	2183.90
2432	Venoge	Ecublens, Les Bois	532040 / 154160	0.0	228.00	686	nivo-pluvial jurassien	9.62	1148.17	332.99	539.44
2433	Aubonne	Allaman, Le Coulet	520720 / 147410	0.0	105.00	952	nivo-pluvial jurassien	8.21	1444.62	340.37	1587.98
2434	Dünnern	Olten, Hammermühle	634330 / 244480	0.0	234.00	711	pluvial jurassien	8.50	1210.83	334.79	437.11
2436	Chli Schliere	Alpnach, Chilch Erli	663800 / 199570	0.0	21.60	1345	nivo-pluvial préalpin	5.96	1876.39	271.62	966.44
2437	Parimbot	Ecublens, Eschiens	552060 / 161650	0.0	6.92	716	pluvial jurassien	9.50	1182.10	311.77	717260.34
2450	Wigger	Zofingen	637580 / 237080	0.0	366.00	656	pluvial inférieur	8.80	1182.26	328.47	461.60
2457	Aare	Ringgenberg, Goldswil	633730 / 171510	12.1	1138.00	1951	mixed regime (>500km^2)	2.47	1761.77	138.78	1715.63
2458	Seyon	Valangin	559370 / 206810	0.0	112.00	978	nivo-pluvial jurassien	7.54	1292.29	350.00	214.91
2461	Magliasina	Magliaso, Ponte	711620 / 93290	0.0	34.40	926	pluvio-nival méridional	8.91	1938.53	357.01	1117.90
2468	Sitter	St. Gallen, Bruggen/Au	742540 / 253230	0.0	261.00	1042	nivo-pluvial préalpin	7.22	1722.67	343.20	1208.91
2471	Murg	Murgenthal, Walliswil	629340 / 233555	0.0	183.00	653	pluvial inférieur	8.60	1191.59	333.34	556.21
2473	Rhein	Diepoldsau, Rietbrücke	766280 / 250360	0.6	6299.00	1771	mixed regime (>500km^2)	3.17	1327.19	193.52	1163.69
2474	Calancasca	Buseno	729440 / 127180	0.2	121.00	1931	nival méridional	2.65	1673.78	227.69	1113.07
2475	Maggia	Bignasco, Ponte nuovo	690040 / 132550	0.9	316.00	1879	nival méridional	3.67	1939.62	187.23	415.67
2477	Lorze	Zug, Letzi	680600 / 226070	0.0	100.00	818	downstream lake	8.15	1560.20	295.43	925.59
2478	Birse	Soyhières, Bois du Treuil	596780 / 249070	0.0	569.00	805	nivo-pluvial jurassien	8.06	1265.37	334.76	580.60
2480	Areuse	Boudry	554350 / 199940	0.0	378.00	1077	nivo-pluvial jurassien	6.27	1464.77	347.62	906.66
2481	Engelberger Aa	Buochs, Flugplatz	673555 / 202870	2.5	228.00	1609	b-glacio-nival	4.30	1693.58	196.94	1705.73
2485	Allaine	Boncourt, Frontière	567830 / 261200	0.0	212.00	562	pluvial jurassien	9.54	1108.82	343.10	464.95
2486	Veveyse	Vevey, Copet	554675 / 146565	0.0	64.50	1098	nivo-pluvial préalpin	7.37	1497.99	307.89	955.88
2487	Kleine Emme	Werthenstein, Chappelboden	647870 / 209510	0.0	311.00	1167	nivo-pluvial préalpin	6.61	1695.60	279.00	1095.28
2488	Simme	Latterbach	610680 / 167840	1.5	563.00	1594	nival de transition	4.79	1506.56	225.96	342.76
2490	Allondon	Dardagny, Les Granges	488880 / 119460	0.0	119.00	760	nivo-pluvial jurassien	10.64	1372.60	326.46	854.70
2491	Schächen	Bürglen, Galgenwäldli, nur Hauptstation	692480 / 191800	1.5	108.00	1728	nivo-glaciaire	3.55	1854.00	188.20	1397.24
2493	Promenthouse	Gland, Route Suisse	510080 / 140080	0.0	120.00	1027	nivo-pluvial jurassien	7.73	1577.83	338.92	430.80
2494	Ticino	Pollegio, Campagna	716120 / 135330	0.2	444.00	1796	nival méridional	3.85	1710.59	173.72	1438.77
2497	Luthern	Nebikon	640560 / 226740	0.0	105.00	749	pluvial inférieur	8.31	1268.69	334.87	429.99
2498	Glenner	Castrisch	735330 / 181790	1.1	381.00	2022	nivo-glaciaire	2.11	1307.02	168.97	734.66
2500	Worble	Ittigen	603005 / 202455	0.0	67.10	666	pluvial inférieur	8.72	1174.24	317.69	475.10
2602	Rhein	Domat/Ems	753890 / 189370	0.9	3229.00	2013	mixed regime (>500km^2)	2.19	1277.76	168.28	1122.84
2603	Ilfis	Langnau	627320 / 198600	0.0	187.00	1039	nivo-pluvial préalpin	7.05	1619.21	318.70	882.14
2604	Biber	Biberbrugg	697240 / 223280	0.0	31.90	1003	nivo-pluvial préalpin	7.00	1789.15	287.37	1085.77
2605	Verzasca	Lavertezzo, Campiòi	708420 / 122920	0.0	185.00	1651	nivo-pluvial méridional	5.11	2013.18	260.43	1846.37
2606	Rhône	Genève, Halle de l'Ile	499890 / 117850	7.2	8000.00	1658	mixed regime (>500km^2)	4.08	1286.36	204.12	996.79
2607	Goneri	Oberwald	670467 / 153932	4.0	38.50	2383	b-glacio-nival	0.07	1976.16	121.82	2011.50
2608	Sellenbodenbach	Neuenkirch	658530 / 218290	0.0	10.40	608	pluvial inférieur	9.33	1193.97	305.25	637.12

T = Temperature, P = Precipitation, E = Evaporation, Q = Streamflow/Runoff

Table B10. Characteristics of all 184 catchments in the HYD-RESPONSES dataset (Part 5/5).

Catchment	Water name	Place	Lon / Lat	Glaciation	Area	Avg. Height	Regime Type	Yearly Avg. T	Yearly P	Yearly E	Yearly Q
			EPSG:21781	%	km^2	m asl		°C	mm	mm	mm
2609	Alp	Einsiedeln	698640 / 223020	0.0	46.70	1157	nivo-pluvial préalpin	6.35	1939.89	279.80	1459.07
2610	Scheulte	Vicques	599485 / 244150	0.0	72.70	792	nivo-pluvial jurassien	8.20	1260.81	333.18	635.64
2612	Riale di Pincascia	Lavertezzo	708060 / 123950	0.0	44.50	1705	nivo-pluvial méridional	4.89	1978.57	277.65	1911.93
2617	Rom	Müstair	830800 / 168700	0.0	128.00	2184	nival alpin	1.03	844.68	158.63	577.25
2620	Mera	Soglio	760770 / 133450	7.4	177.00	2173	b-glacio-nival	0.95	1333.42	191.09	361.81
2629	Vedeggio	Agno, stazione principale	714110 / 95680	0.0	99.90	921	pluvio-nival méridional	8.87	1904.96	334.27	656.21
2630	Sionne	Sion	594400 / 119900	0.0	27.60	1577	nival alpin	5.35	1355.03	186.55	224.58
2631	Hinterrhein	Hinterrhein, Schiessplatz	733706 / 153945	9.1	41.50	2430	a-glacio-nival	-0.38	1704.08	164.40	799.71
2634	Kleine Emme	Emmen	663700 / 213630	0.0	478.00	1054	nivo-pluvial préalpin	7.15	1610.61	284.17	994.18
2635	Grossbach	Einsiedeln, Gross	700710 / 218125	0.0	8.95	1283	nivo-pluvial préalpin	5.90	1952.69	299.45	1387.71
2640	Sorne	Delémont, Pré-Guillaume	593380 / 245940	0.0	214.00	779	nivo-pluvial jurassien	8.20	1233.60	335.80	603.72
2646	Kander	Emdthal	617790 / 168400	5.1	487.00	1860	b-glacio-nival	3.38	1486.71	167.85	1305.44

T = Temperature, P = Precipitation, E = Evaporation, Q = Streamflow/Runoff

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