Different drought types and the spatial variability in their hazard, impact, and propagation characteristics

Erik Tijdeman[1], Veit Blauhut[2], Michael Stoelzle[2], Lucas Menzel[1], and Kerstin Stahl[2]

[1] Professorship of Hydrology and Climatology, Institute of Geography, Heidelberg University, Heidelberg, Germany
[2] Faculty of Environment and Natural Resources, University of Freiburg, Freiburg, Germany

Correspondence to: michael.stoelzle@hydro.uni-freiburg.de

Abstract

Droughts often have a severe impact on environment, society, and economy. ~~Only a multifaceted assessment of such droughts and their impacts can provide insights in t~~The variables and scales that are relevant to understand the impact of drought ~~for drought management. M~~motivated ~~by~~ this ~~aim~~study, ~~we~~ which compared hazard and propagation characteristics as well as impacts of major droughts between 1990-2019 in Southwestern Germany. We bring together high-resolution datasets of air temperature, precipitation, soil moisture simulations, streamflow and groundwater level observations, as well as text-based information on drought impacts. Various drought characteristics were derived from the hydrometeorological and drought impact time series and compared across variables and spatial scales. Results revealed different drought types sharing similar hazard and impact characteristics. The most severe drought type identified is an intense multi-seasonal drought type peaking in summer, i.e. the events in 2003, 2015 and 2018. This drought type appeared in all domains of the hydrological cycle and coincided with high air temperatures, causing a high number and variability of drought impacts. The regional average drought signals of this drought type exhibit typical drought propagation characteristics such as a time lag between meteorological and hydrological drought, whereas propagation characteristics of local drought signals are variable in space. This spatial variability in drought hazard increased when droughts propagated through the hydrological cycle, causing distinct differences among variables, and regional average and local drought information. Accordingly, single variable or regional average drought information is ~~considered to be~~ not sufficient to fully explain the variety of drought impacts that occurred~~. In addition to l~~ supporting the conclusion that in regions as diverse as the case study presented here, large-scale drought monitoring~~, drought management~~ needs to ~~consider~~ be complemented by local drought information to assess the multifaceted impact of drought ~~from different hydrometeorological variables and could be type based~~.

1 Introduction

The Central and Northern European drought and heatwave of 2018 revealed once more the large spatial-temporal footprint and severe impacts of this natural hazard (e.g. Bakke et al., 2020; Brunner et al., 2019; Schuldt et al., 2020). Similar or worse

episodes are expected to occur more often in the future, given the increasing atmospheric water demand and human pressure on fresh-water resources (e.g. Samaniego et al., 2018; Wanders and Wada, 2015). This prospect raises the importance of short- and long-term drought management to better cope with both ongoing drought as well as to be better prepared for future drought episodes (Wilhite et al., 2019). Good planning for drought requires information about the different components of drought

35 risk: hazard, impacts, exposure, and vulnerability. The analysis of past droughts at different scales can provide this information. On the one hand, locally relevant drought management benefits from detailed information, which considers different hydrometeorological variables and drought related impacts and their spatiotemporal variability (e.g., Van Lanen et al., 2016). On the other hand, higher-level administrative decision-making often requires drought information in a more generalized form, e.g., indexed information aggregated to administrative regions indicating whether there is drought or not. Generalizing drought

40 information simplifies its interpretation but may come at the cost of a loss of information, as the hazard and its impacts may be highly variable in space and time (Stahl et al., 2016).

The multifaceted nature of drought results in the impracticability of a unique drought definition (Lloyd-Hughes, 2014). From a hazard perspective, drought is often defined as a below normal hydrometeorological anomaly, where the normal depends on space, time, and the variable of interest (e.g., Tallaksen and Van Lanen, 2004). Accordingly, a wealth of drought hazard indices

45 has been developed to express the flux or state of a certain domain of the hydrological cycle, or the combined states and fluxes of multiple domains, as anomaly (e.g. Hao and Singh, 2015; Zargar et al., 2011). From an impact perspective, below normal anomalies only become droughts when they have the potential to cause drought impacts, i.e., "negative consequences of drought for environment, society or economy" (Blauhut et al., 2015).

The above-described definitions of drought form the basis of many drought-related studies, which often focus on a specific

50 type of drought such as meteorological, agricultural, hydrological, or socio-economic drought (Wilhite and Glantz 1985). Laaha et al. (2017) argue that our understanding of drought would benefit from a more holistic study of drought phenomena, because specific drought impacts relate to droughts in certain domains of the hydrological cycle. In addition, the co-occurrence of drought in different domains of the hydrological cycle may worsen drought impacts. For example, agricultural drought impacts caused by low soil moisture can be aggravated by co-occurring streamflow droughts that limit or prohibit withdrawals

55 of surface water for irrigation. A holistic view on drought and its impacts might further benefit the consideration of compound hazards preceding or co-occurring with drought, e.g., heat waves, as these can further worsen drought impacts (Zscheischler et al. 2020). Finally, a temporal clustering of drought years might be worth the consideration. A second drought year in a row, as observed for meteorological drought for Central and Northern Europe in 2018-2019 as well as for various hydrological droughts in the UK (resp. ectively Hari et al., 2020; Kendon et al., 2013), prevents recovery and might have an even larger

60 impact on already weakened systems.

The connection of different drought types through so-called "drought propagation" (e.g., Changnon, 1987; Van Loon, 2015) further justifies the need for a holistic drought assessment. Drought propagation is a well-established concept on the catchment scale (e.g. Van Loon, 2015). Rainfall deficits enhanced by meteorological conditions that favor high evapotranspiration propagate to deficits in root zone soil moisture followed by recession and ultimately deficits in river flow and groundwater.

65 Catchment scale drought propagation exhibits various typical ~~characteristics, including~~ characteristics. One of these characteristics is the order of appearance of drought in different domains of the hydrometeorological cycle and associated time-lag, where meteorological droughts are followed by soil moisture droughts, which are eventually followed by streamflow droughts and finally groundwater droughts (e.g. Changnon, 1987; Peters, 2003). Another propagation characteristic ~~and associated time-lag as well as the~~ is the attenuation of the propagating hydrometeorological signal and associated lengthening.

70 The time series of hydrometeorological variables generally get smoother the further they propagate through the hydrological cycle, e.g., a monthly precipitation time series is often much more variable than a monthly groundwater time series. Accordingly, drought periods in less variable drought signals tend to be longer (lengthening), i.e., meteorological droughts are generally shorter than soil moisture droughts, which are shorter than streamflow droughts and groundwater droughts (e.g. Hisdal and Tallaksen, 2000, Van Loon and Van Lanen, 2012; Van Loon et al., 2014). Regional scale drought propagation is

75 a less well-established concept. On this regional scale, drought propagation may not happen uniformly, as the propagating drought signal is affected by meteorological variability and further modified by geological differences in surface and sub-surface characteristics. Differences in meteorological conditions are driven by both larger- and smaller-scale atmospheric processes. On the larger (continental) scale, atmospheric circulation patterns and teleconnections play an important role, e.g., by blocking wet weather systems (e.g., Ionita et al., 2017; Toreti et al., 2019). On the smaller regional scale, differences in

80 meteorological conditions appear due to the occurrence of local rainstorms or differences in topography and land cover that affect energy balance terms and hence air temperature, evapotranspiration, and meso-scale wind systems. Differences in subsurface soil and geological characteristics may also exert a strong control on regional drought propagation, as these affect the total amount of storage and thereby influence how well meteorological dry spells can be buffered (e.g., Barker et al., 2016; Bloomfield and Marchant, 2013; Heudorfer et al., 2019; Stoelzle et al., 2014).

85 The variable occurrence and severity of drought in space, time, and among different domains of the hydrological cycle may be a major precursor of when and where different types of drought impacts occur. However, drought impact occurrence is not solely related to the hazard as drought impacts are "symptoms of vulnerability" (Knutson et al, 1998). Thus, an indicated drought hazard might not necessarily lead to an impact, as the exposed system also has to be vulnerable to the hazard. For example, certain drought impacts might not occur because of appropriate mitigation measures in place. In addition, the

90 considered spatially aggregated or single variable drought hazard information might not be representative for a specific type of local drought impacts.

It is the question whether a holistic assessment of droughts enhances the understanding of the type and amount of drought impacts that occur during certain types of drought events~~. as this~~If this is the case, it ~~could~~ would highlight the need for multivariate and drought type specific management. It is further the question whether concepts of catchments scale drought

95 propagation also exist on the regional scale, as these concepts could serve as guiding principles for regional drought management. For both these questions, spatial scale, i.e., whether local or regional aggregated drought information is investigated, may play an important role. What is the unique value of different drought information sources at different scales? And is there a benefit of using local over spatially aggregated drought information in a regional drought management context?

These questions have not yet been systematically explored as often data is lacking for such comprehensive analyses. To fill this gap, this study aims to shed more light on regional drought (propagation) characteristics with a multivariate view on recent droughts to enhance the understanding of drought impacts. Focusing on Southwestern Germany as a case study, we aim to:

1. identify drought episodes for the period 1990-2019 and investigate if these episodes can be grouped into different types with similar hazard characteristics and impacts,
2. evaluate the propagation of drought over the region for the most prominent (highest impact) type's drought episodes, and
3. assess the agreement among drought hazard and impact information across variables and scales.

2 Data and methods

2.1 Study area

Baden-Württemberg is the most Southwestern federal state of Germany (Fig. 1). According to the Nomenclature of Units for Territorial Statistics (NUTS) of the European Union, Baden-Württemberg is a NUTS-1 region, which is separated into four NUTS-2 regions: Stuttgart, Karlsruhe, Freiburg and Tübingen. In this case, these NUTS regions also have their own governmental water authorities. Hydrologically important, the river Rhine flows along the southern and western border of the state and provides an important waterway for navigation. The region encompasses both flat and lowland areas such as the Rhine Valley in the West as well as more mountainous regions such as the Black Forest and the Swabian Alb. The topography of the region affects both precipitation (annual average sum between 600 and >2000 mm year^{-1}) and air temperature (annual average between 4.5 and 11.6 °C). The landscapes of Baden-Württemberg are diverse, with agricultural (43%) and urban (7%) areas mostly located in the lower elevated regions and forested (38%) and meadow areas (10%) in the higher elevated regions. Equally diverse are the region's lithological and geological characteristics. A variety of soils with different root-zone depths and water holding capacities exist, ranging from thick loess layers to shallow Leptosols. Below these soils, the geology varies from metamorphic rock to porous limestone and unconsolidated rock, resulting in a hydrogeology with different aquifer types that have different storage capacities (Stoelzle et al., 2015). Aquifers also vary in size, with some larger porous aquifers in e.g., the Rhine Valley and smaller valley fill aquifers among fractured sedimentary or crystalline bedrock in e.g., the Black Forest.

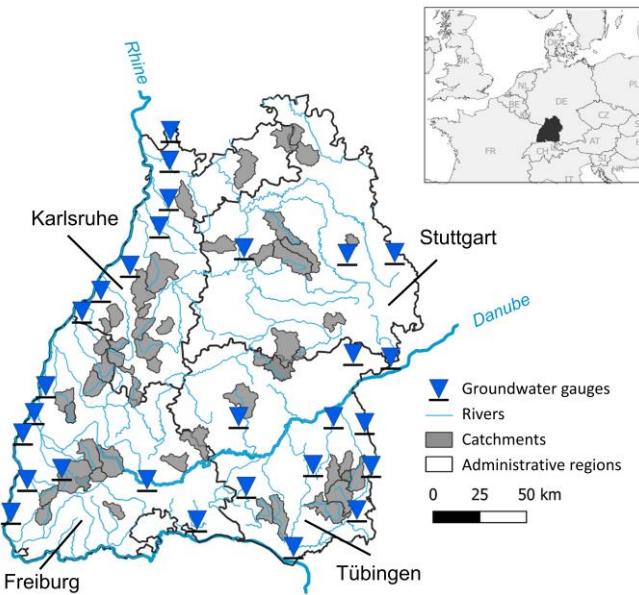



Figure 1. Data basis for the study in the federal state of Baden-Württemberg (NUTS-1) with its four different administrative regions
125   **(NUTS-2).**

2.2 Data and temporal aggregation

The study assembled multiple variables for the region: air temperature, precipitation, simulated soil moisture and observed streamflow and groundwater levels. These daily hydrometeorological observations and simulations for the period 1990-2019 stem from different sources. Air temperature data come from the interpolation of ~~own~~ climate station ~~interpolations~~ data over
130   a 1-km resolution grid covering Baden-Württemberg (Tijdeman and Menzel 2021a). Precipitation data stem from the gridded REGNIE data product (Rauthe et al., 2013) and was sourced from the Climate Data Center of the German Weather Services (DWD, 2020). Gridded simulated soil moisture data over a 1-km resolution grid were obtained from Tijdeman and Menzel (2021b). These soil moisture simulations were derived using the TRAIN (TRAnspiration and Interception) model; a physical based model simulating fluxes and state variables on the soil-water-atmosphere interface. The performance of the TRAIN
135   model was evaluated against soil moisture and river flow observations in Tijdeman and Menzel (2021a) and various studies mentioned therein. ~~(described in Tijdeman and Menzel 2021a) and were obtained from Tijdeman and Menzel (2021b).~~ Daily streamflow observations at gauging stations of 54 catchments with flow unaffected by major human disturbances at the timescale relevant for our analyses were provided by the Ministry of Environment of Baden-Württemberg (LUBW). Groundwater level observations of 43 wells were obtained from the Environmental Data and Maps archive of the LUBW
140   (UDO, 2020). We only considered those groundwater wells currently in use in the reference network of the groundwater storage assessment application of the LUBW (LUBW, 2020), given their relevance for drought monitoring. This does not mean that all groundwater level time series are completely free of human influences. Nevertheless, Baden-Wurttemberg is a

water rich country, and visual inspection of the groundwater level time series did not show any sharp changes associated with changes in (nearby) abstractions.

145 All above-described variables are continuous daily observations, except for groundwater level observations, which are available at (ir)regular intervals, e.g., at certain days of the week. To obtain a common temporal resolution, we aggregated daily hydrometeorological data to a monthly resolution by taking the mean of the daily observations or simulations. This resulted in monthly time series of precipitation (P1), air temperature (T1) simulated soil moisture (SM) streamflow (Q) and groundwater levels (GW) for either each grid cell (P1, T1, SM), catchment (Q), or well (GW). Precipitation was further

150 aggregated to a seasonal and annual resolution (respectively. P3 and P12), one value for each calendar month, to account for both short- and long- term meteorological deficits relevant for e.g., hydrological systems with respectively. a low- and high-buffering capacity of meteorological deficits. Air temperature was also aggregated to an annual resolution (T12), one value for each year, to depict the general changes in climate. For T12, we deliberately looked at a static annual window and not to a moving 12-month window as we do for P12, because the former static approach is more in line with the methods to display

155 changes in annual air temperature. Daily time series of precipitation, air temperature and simulated soil moisture did not contain any missing values, whereas daily time series of streamflow and groundwater level observations contained occasional gaps. In the case of gaps in daily streamflow, a certain month was set to missing whenever five or more days in that month did not have an observation. For groundwater, we used less stringent missing data requirements, given that groundwater levels were not available at a continuous daily resolution. GW in a certain month was set to "missing" when there were no observations in that

160 month. In the end, we only selected the monthly Q and GW level time series with less than 10% of "missing" months, respectively all of the 54 streamflow time series and 28 out of 43 of the groundwater level time series (Fig. 1).

Drought impact information stems from the European Drought Impact report Inventory (EDII; Stahl et al., 2016); a database of textual drought impact information from different sources, e.g., newspaper articles, governmental reports etc. These reports are manually collected, temporally and spatially referenced, and classified to one of 15 impact categories (Blauhut et al., 2015).

165 The timestamp of an entry indicates at a minimum the year of occurrence, but where available also information on the beginning and in some cases the end date of a reported impact. For this study, we considered impacts for the period 2000-2019 (n = 792), given the more limited availability of impact information prior to this date. The EDII has recently been updated for the greater alpine region (Stephan et al., 2021). For the study region of Baden-Württemberg this dataset was also supplemented with additional impact information gathered from questionnaires of a survey of the hydropower sector (Siebert et al. 2021;

170 until 2017) and a survey of the public water supply sector (Blauhut et al. 2020; until 2018). Categorical impact information was grouped by start year, season, where available start month (n = 359), and NUTS-1 and NUTS-2 region. The grouping over NUTS regions was used as these regions coincide with administrative boundaries relevant for decision-making according to the state's Water Act.

2.3 Drought hazard information at different spatial scales

175 We derived drought hazard information from all meteorological and hydrological variables at three spatial scales: local, NUTS-2 average, and NUTS-1 average. Local drought information was derived for each individual grid cell, catchment or well by transferring their time series to anomaly space (percentiles; p) using Weibull plotting positions (Weibull 1939, eq. 1).

$$p_{V,u} = \frac{\text{Rank}(V_u)}{(n+1)} \qquad \text{(eq. 1)}$$

180

Where V is the variable of interest, n the sample size (in this study: $n = 30$) and u the location identifier referring to either a grid cell, catchment, or groundwater well. The rank of V in a specific year and month is relative to historical observations in that month, e.g., P1 for August 2018 compared to P1 for all other Augusts. To derive anomalies from air temperature time series (T1, T12), we used an inverse ranking to make the percentile classification of above normal air temperature comparable

185 with below normal hydrometeorological conditions. The derived percentiles express the historical non-exceedance frequencies of different variables for the considered reference period 1990-2019 (Tijdeman et al., 2020). This reference period differs from the standard reference period of the WMO (e.g., 1961-1990). However, using the period 1990-2019 as reference was preferred given the lower availability of continuous observations for the earlier period.

For the NUTS-1 region as a whole and for the four NUTS-2 sub-regions, regional average drought hazard information was

190 derived by ranking the average of all available local percentile data within the region; for a given hydrometeorological or hydrological variable, the regional average p is then (eq. 2).

$$p_{V,\text{NUTS}} = \frac{\text{Rank}(\overline{p_{V,\text{NUTS}}})}{(n+1)} \qquad \text{(eq. 2)}$$

195 Where NUTS refers to either all grid cells, catchments or wells located within either the NUTS-1 or NUTS-2 region of interest. The rank of $\overline{P_{V,\text{NUTS}}}$ thus compares average local percentile values for a specific NUTS-1 or NUTS-2 region, variable, year, and month to average percentile values for the same NUTS-1 or NUTS-2 region, variable, and month in all other years. The NUTS-1 and NUTS-2 average drought signals were used to identify major drought episodes over the region. These episodes then serve as a starting point to further explore the spatial variation in local drought hazard and impact information during

200 major drought episodes.

The percentile time series of all hydrometeorological variables and scales were classified into five different groups: $p_V \leq 0.1$ (severe drought or much above normal air temperature), $0.1 < p_V \leq 0.25$ (moderate drought or above normal air temperature), $0.25 < p_V < 0.75$ (normal conditions), $0.75 \leq p_V < 0.9$ (moderately wet or below normal air temperatures) and, $p_V \geq 0.9$ (severely wet or much below normal air temperatures). For ease of notation, we move the location identifier outside of the subscript and

205 use the variable abbreviation to refer to its percentiles where applicable, e.g., local P1 when referring to monthly precipitation

percentiles at individual locations (cells) or NUTS-1 average P1 when referring to regional average monthly precipitation percentiles.

2.4 ~~Typing hazard characteristics and impacts~~ Different drought episode types and their characteristics and impacts

We characterized past drought episodes over the period of 1990-2019 and the impacts of recent events from 2000-2019. First, we identified drought episodes from the NUTS-1 average percentile time series of all hydrometeorological variables. These episodes were classified into different types considering timing and length of the episode as well as the affected domains of the hydrological cycle. These aspects of drought episodes were considered as we hypothesize that they influence the type and amount of related drought impacts that occur.

In a next step, we quantified whether different types of drought episodes also differ in hazard and impact characteristics. We first divided the NUTS-1 average percentile time series (p) of all hydrometeorological variables (v) into drought ($p_V \leq 0.25$) and non-drought events ($p_V > 0.25$). We then derived the duration (D, months) and severity (S, -) of each event j (eq. 3 & 4, respectively):

$$D_{\mathrm{pv,total}}[\,j\,] = \sum_{t=1}^{D_{\mathrm{P_V}}(\mathrm{j})} 1 \quad (\mathrm{eq.}\,3)$$

$$S_{\mathrm{pv,total}}[\,j\,] = \sum_{t=1}^{D_{\mathrm{P_V}}(\mathrm{j})} 0.25 - p_V(\mathrm{t}) \quad (\mathrm{eq.}\,4)$$

S hence incorporates both the duration of the event and the deviation from the threshold but has no physical relation to water quantity. S should be seen as a relative metric that only enables comparison among other variables' severity values. With regard to drought impact characteristics, we considered the number and categories that happened during the different types of drought episodes. The number of impacts gives an indication of the severity and perception of the drought events, whereas impact categories provide insight in the (diversity of) affected sectors.

2.5 Regional drought propagation

We introduce regional drought propagation, i.e., the propagation of drought through the hydrological cycle in space and time over a larger region. Focusing on the most prominent and also highly impacted drought episodes, it was investigated whether regional drought propagation exhibited typical characteristics known from catchment scale drought propagation such as ordering, time-lag, and lengthening. We first inspected whether these drought propagation characteristics were present in NUTS-1 average, NUTS-2 average, and local percentile time series. We then quantified the initiation time (I, months) and maximum duration (M, months) of each drought episode from both NUTS-1 average and all local percentile time series of all hydrometeorological variables (p_V), where:

235   - I_{pv} is the initiation time, i.e., the time between an arbitrary set starting point (e.g. start of drought development) and the first time the percentile time series reached drought ($p_V \leq 0.25$), and

-    M_{pv} is the maximum time (months) the percentile time series was continuously in drought ($p_V \leq 0.25$) during the drought episode.

From I_{pv} and M_{pv} of all local percentile time series of each variable, we derived the 5th and 95th quantile to differentiate

240   between quick ($I_{pv,Quick}$) and slow ($I_{pv,Slow}$) developing drought conditions as well as short ($M_{pv,Short}$) and prolonged ($M_{pv,Long}$) local drought conditions. Quickly developing indicates the time it took for the first 5% of local drought signals to reach drought conditions and slowly developing to the time it took for 95% of the local drought signals to reach drought conditions.

The ordering and difference of I_{pv} among variables is indicative for respectively the ordering and time lag between drought

245   events appearing in different domains of the hydrological cycle. An amplification in M when drought events propagate through the hydrological cycle is indicative of lengthening. We hypothesize that the order of appearance, time lag, and lengthening of drought during drought episodes are generally visible in the NUTS-1 average and local drought signals such that:

-    The precipitation drought develops before the soil moisture drought, the soil moisture drought develops before the streamflow drought, and the streamflow drought develops before groundwater droughts ($I_{P1} \leq I_{SM} \leq I_Q \leq I_{GW}$)

250   -   The precipitation drought is shorter than the soil moisture drought, the soil moisture drought is shorter than the streamflow drought, and the streamflow drought is shorter than groundwater drought ($M_{P1} \leq M_{SM} \leq M_Q \leq M_{GW}$)

However, we also speculate that regional differences in climate, soil, catchment, and aquifer characteristics can modify the order of appearance and lengthening of local drought conditions over a larger region such that:

-    Slower developing local soil moisture drought conditions develop later than more quickly developing local
255   streamflow drought conditions and slower developing local streamflow drought conditions develop later than quickly developing local groundwater drought conditions ($I_{SM,Slow} \geq I_{Q,Quick}$; $I_{Q,Slow} \geq I_{GW,Quick}$).

-    The most prolonged local soil moisture drought conditions exceed the duration of the shortest local streamflow drought conditions and the most prolonged streamflow drought conditions exceed the duration of the shortest local groundwater drought conditions ($M_{SM,Long} \geq M_{Q,Short}$; $M_{Q,Long} \geq M_{GW,Short}$).

260   **2.6 Agreement among different drought information sources**

The usefulness of different levels and sources of drought information was assessed based on their agreement (A; between 0 and 100). In this study, we assess the agreement between NUTS-1 average drought signals (P_{V1}) with drought signals of other variables (P_{V2}) at different scales and locations (u). Depending on the scale (NUTS-1, NUTS-2, local), u (u_{total}) refers to either the (total number of) NUTS regions, local grid cells, rivers, or groundwater wells. A was derived by dividing the number

265   of times two drought information sources (P_{V1} and P_{V2}) were in drought (eq. 5)

$$A_{numerator} = \sum_{t=1}^{T} \sum_{u=1}^{u_{total}} \begin{cases} 1 & if \quad P_{V1,t} < 0.25 \; and \; P_{V2,t,u} < 0.25 \\ 0 & otherwise \end{cases} \quad \text{(eq. 5)}$$

by the total number of times the second drought information source (P_{V2}) was in drought.

$$A_{denominator} = \sum_{t=1}^{T} \sum_{u=1}^{u_{total}} \begin{cases} 1 & if \quad P_{V2,t,u} < 0.25 \\ 0 & otherwise \end{cases} \quad \text{(eq. 6)}$$

such that A is (eq. 7):

$$A = \frac{A_{numerator}}{A_{denominator}} * 100 \quad \text{(eq. 7)}$$

Where t=1 indicates the start of the drought episode of interest and T the end of this period. For this analysis, one prominent drought episode was chosen, so that t=1 indicates the first month of this episode and T the last month. The lower A among two different drought information sources, the higher the unique value of each individual source, and vice versa, the higher A, the lower the unique value of each individual source. For this analysis, one prominent drought episode was chosen. In addition, we investigated the agreement between drought impact occurrence and drought hazard information of different variables at different scales and locations by replacing the conditional statement $P_{V1,t} < 0.25$ in equation 5 with the condition that at least one impact has started in month t. For this episode, we derived A by comparing the fraction of cases each individual drought (p_V<0.25) information source (different variables and different scales) was in drought for all months in which:
- The NUTS-1 average percentile time series of P1, SM, Q or GW was in drought, or
- at least one drought impact started.

To Finally, gain more insight in the variability in hydrometeorological conditions beyond agreement metric A, we also derived the percentile class distributions of the NUTS-1, NUTS-2, and local percentile time series for all months in which the NUTS-1 average percentile time series was in drought or all months in which at least one drought impact report started.

3 Results

3.1 Different drought episode types and their characteristics and impacts

NUTS-1 average percentile time series of different variables reveal several past drought episodes in the study region (Fig. 2). Based on common characteristics, these episodes can be grouped into three distinct types and two combinations of them:
- intense multi-seasonal drought episodes peaking in summer as in 2003, 2015 and 2018 (Type I),
- long-term and less intense drought episodes in e.g., the early 1990s (Type II),
- short-term episodes as in e.g., 2011 & 2014 (Type III),
- Type I episodes that transition in Type II episodes as in e.g., 2004 & 2019 (referred to as Type I→II) and,
- Type III episodes that coincide with Type II episodes as in e.g., 1998 (referred to as Type III+II).

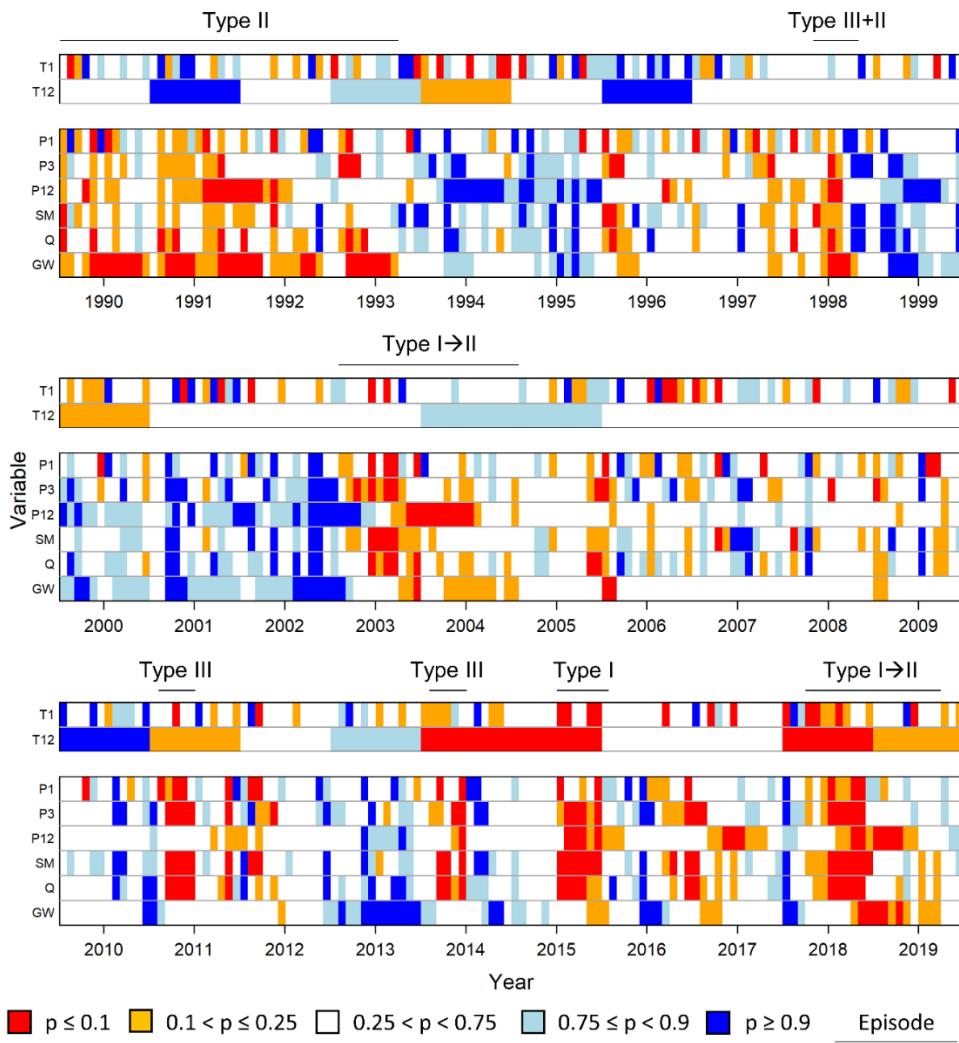



Figure 2. Drought episode types identified from NUTS-1 average classified percentile (p) time series of monthly and annual air temperature (resp. ectively T1 and T12), precipitation accumulated over different periods (P1, P3 and P12), simulated soil moisture (SM) and observed streamflow (Q) and groundwater levels (GW). Tick marks on the x-axis showing the years were placed on the first of July of that year.

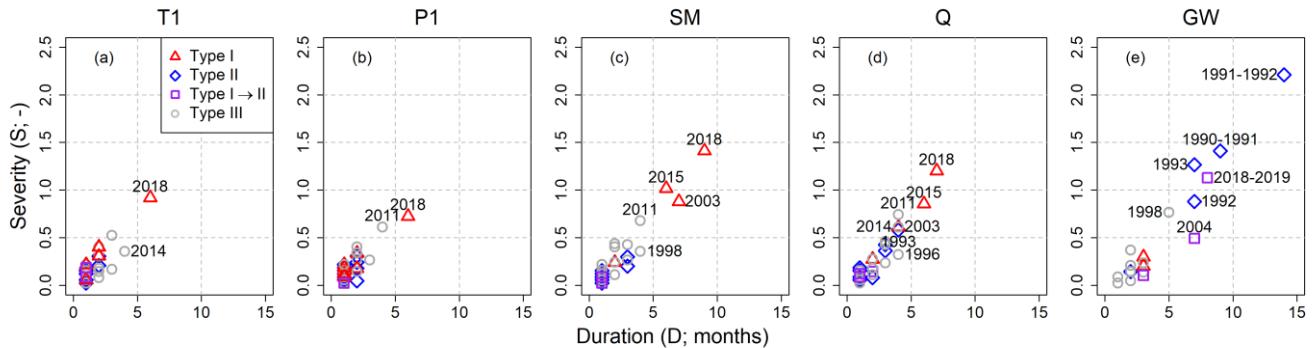
Type I episodes showed persistent meteorological dry spells over the growing season according to NUTS-1 average P1 and P3 that caused severe and prolonged deficits in NUTS-1 average SM and Q. Type II and Type I→II episodes were characterized by long-term meteorological dryness as indicated by NUTS-1 average P12 and were associated with persistent below normal NUTS-1 average GW, occasionally interrupted by some wetter months. The shorter Type III episodes caused below normal NUTS-1 average SM and Q. However, the impact of these shorter dry spells on NUTS-1 average GW varied among the years and depended on the initial conditions of the groundwater systems at the start of the drought. The Type III+II episode of 1998

was preceded by a period of long-term dryness and coincided with below normal NUTS-1 average GW, whereas the Type III episodes of 2011 and 2014 happened after a relatively wet period and did not cause below normal NUTS-1 average GW.

310 The temperature setting in which these different types of drought occurred varied among drought years. There is a general increasing trend towards higher annual air temperatures and NUTS-1 average T12 of four out of the last six years (2014-2019) was warmer than average (relative to the period 1990-2019). Nevertheless, there can be a lot of within year monthly variability according to NUTS-1 average T1. Type I episodes all coincided with at least a few months of extremely high air temperatures. On the other hand, the Type II episode of the early 1990s happened in a relatively cold setting. Air temperature was below

315 normal for the Type III+II episode that occurred in 1998, whereas the Type III episodes in 2011 and 2014 coincided with one or a few warmer months.

The duration (*D*) and severity (*S*) of past drought episodes derived from NUTS-1 average percentile time series vary among variables and drought types (Fig. 3). *D* and *S* of meteorological dry spells was generally low (Fig. 3a-b). A notable exception was the meteorological drought of 2018 with D_{T1} and D_{P1} of six months. For Type I droughts, *D* and *S* increased when droughts

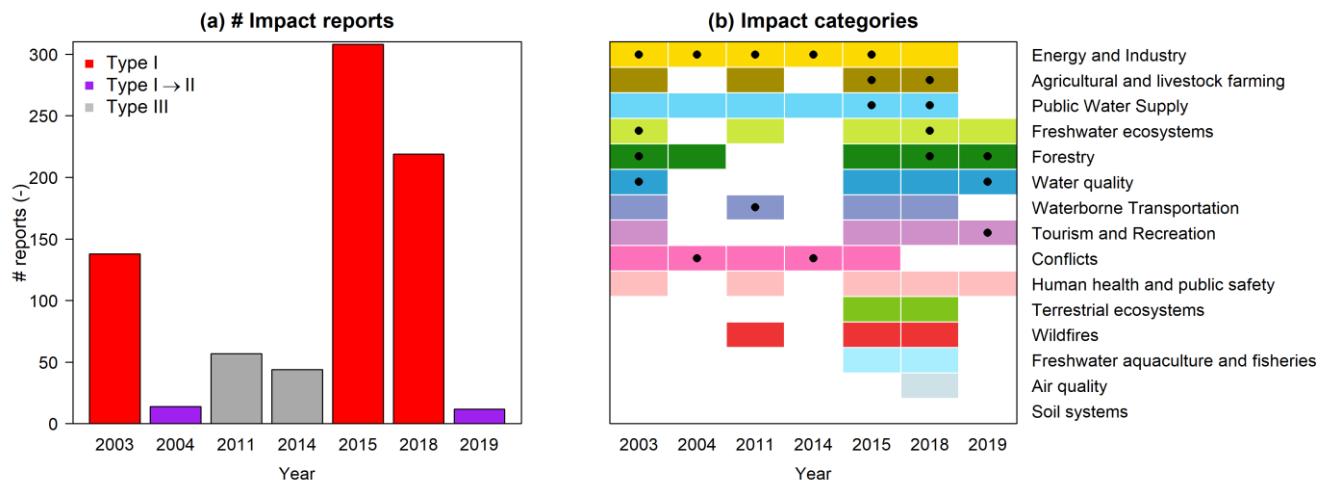
320 propagated from precipitation to soil moisture and streamflow (Fig. 3c-d). Such an amplification in drought characteristics was not visible for the other drought types. On the other hand, the D_{GW} and S_{GW} of type II episodes were exceptional (Fig. 3e). For this drought type, multiple shorter meteorological drought events coincided with prolonged periods of groundwater drought. D_{GW} and S_{GW} were also high after part of the Type I drought episodes (2004, 2018-2019).

325 **Figure 3. Duration (*D*) and severity (*S*) of hydrometeorological drought events during different types of drought episodes derived from monthly NUTS-1 average percentile time series with (a) air temperature (T1) (b) precipitation (P1), (c) simulated soil moisture (SM), (d) streamflow (Q), and (e) groundwater (GW). Events with a duration ≥ 4 months are labeled.**

The total number of reported drought impacts varied among drought years as well as drought types (Fig. 4a). Most impacts were reported for the Type I episodes, lesser for the Type III episodes and least for the Type I→II episodes. The categorical

330 composition of drought impacts revealed that the largest shares of impacts related to the energy and industry sector, agriculture and livestock farming, public water supply, and freshwater ecosystems (Fig. 4b). The most frequently reported impact type in the category of energy and industry was reduced hydropower production. Agricultural impacts include the reduction in harvest quantity and quality or the restriction in irrigation. Public water supply impacts are often related to water use restrictions or

the need to allocate water from other sources, whereas freshwater ecosystem impacts e.g., related to fish die-off. In addition

335 to the more commonly reported impacts, a large variety of other impacted sectors became visible, including forestry, water quality, waterborne transportation, tourism and recreation, and drought related conflicts. The categorical distribution of reported impacts for Baden-Württemberg further reveals drought-type specific differences (Fig. 4b). In general, the highest diversity in impact categories was reported for the Type I drought episodes, whereas drought impacts of Type I→II and Type III episodes were less diverse. A comparison by years further shows event-specific differences. In 2015 and 2018, a relatively

340 large share of impacts on agriculture were reported. The drought of 2011 was dominated by impacts attributed to hydrological drought, i.e., anomalies in streamflow and groundwater levels, such as energy and industry, waterborne transportation, and public water supply. However, no impacts on forestry and agriculture were observed. In contrast, the year 2019 showed a large share of forestry related issues, especially related to the dieback of spruce and bark beetle infestations.

345 **Figure 4. Reported impacts of drought episodes from 2000 to 2019. (a) Annual number of impacts in the European Drought Impact Inventory by drought type, and (b) reported drought impact categories for different drought years, sorted descending by total number of reports. Black dots indicate impact categories that account for more than 10% of all impact reports in a specific year.**

3.2 Regional drought propagation characteristics of prominent Type I and Type I→II episodes

A similarity for the prominent Type I and Type I→II drought episodes is their common start from relatively wet initial

350 conditions in winter (Fig. 5). From that point, several months with below normal P clustered together in prolonged periods of below normal SM and Q, and eventually reached a state of below normal GW. Another similarity is that local drought signals become more variable in space when propagating through the hydrological cycle, as is shown by the increasing ranges in local drought conditions. A further similarity among prominent drought years is the occurrence of some relatively wet months in the following winter. In 2003 and 2018, these single wet months generally recovered drought for SM and Q, but only had a

355 small impact on GW, which often stayed low or continued to decline. The wet January in 2004 was not sufficient for a full recovery of GW, which meant that part of the groundwater levels dropped again to below normal conditions that would persist for the remainder of 2004 and the first part of 2005. The same was observed for the drought starting in 2018; some relatively

wet winter months had little effect on most of the groundwater levels, which stayed below normal throughout the year 2019. On the other hand, the drought of 2015 was followed by multiple wet winter and spring months, which alleviated drought conditions by the start of the summer of 2016.

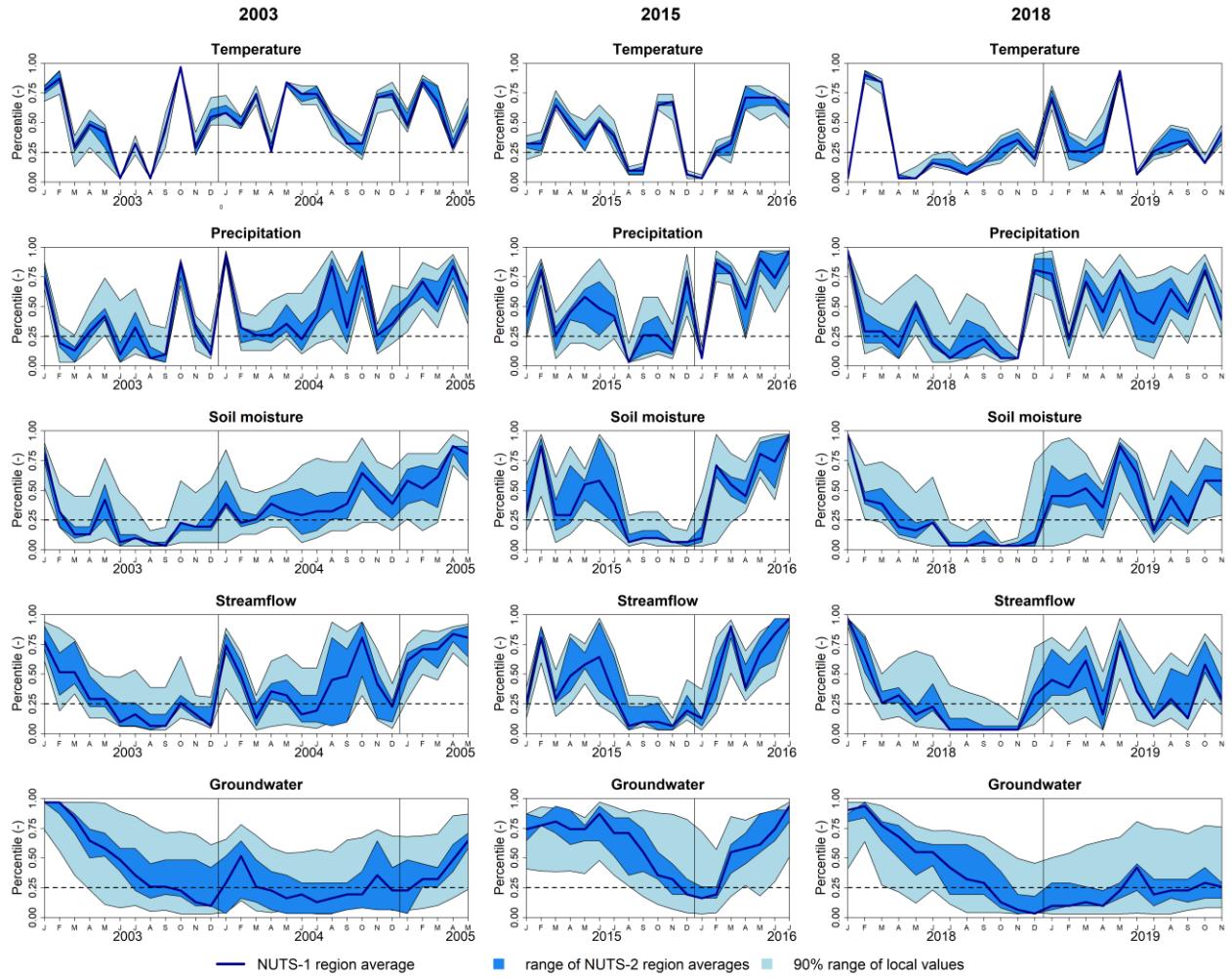



Figure 5. Spatiotemporal propagation of prominent droughts (Type I and Type I→II) according to (ranges in) NUTS-1, NUTS-2 and local percentile timeseries relative to the reference period 1990-2019 (Sect. 2.4). Dashed line indicates the moderate drought or above normal air temperature threshold.

Various typical drought propagation characteristics, i.e., ordering, time-lag, and lengthening, can be recognized from the (ranges in) NUTS-1, NUTS-2, and local percentile time series (Fig. 5). The occurrence of these drought propagation characteristics is partly confirmed when comparing initiation time (I_V) and maximum duration (M_V) among the different variables (Fig. 6). For all three drought episodes, the hypothesized ordering and time lag is visible in NUTS-1 I_V and local $I_{V,Quick}$ (Fig. 6a; i.e., $I_{P1} \leq I_{SM} \leq I_Q \leq I_{GW}$, e.g., for 2003: $2 \leq 3 \leq 6 \leq 10$ months). Local $I_{V,Slow}$ shows this expected ordering for P, SM and Q. However, not all local GW observations reached drought during the drought episodes of 2003 and 2015. The

lengthening of drought is visible when P1 droughts propagate to SM droughts ($M_{P1} \leq M_{SM}$). However, M_Q is occasionally lower than M_{SM} (e.g., 2003) and M_{GW} is occasionally lower than M_Q (e.g., 2015; Fig. 6b).

The ranges in Figure 5 indicate that local drought conditions within a region according to the same variable may vary strongly at one moment in time. The variation also suggests that the temporal sequencing of different drought types was not uniform over a larger region. This is confirmed when comparing local $I_{V,Quick}$ and $I_{V,Slow}$ (diagonal arrows in Fig. 6a). For example, GW of some responsive aquifers reached below normal conditions prior to Q of some less responsive catchments ($I_{Q,Slow} \geq I_{GW,Qucik}$ e.g. for 2003: $12 \geq 4$ months). This is also confirmed when comparing $M_{V,Short}$ and $M_{V,Long}$. For example, the longest time local SM was continuously in drought exceeds the shortest time local Q was continuously in drought ($M_{SM,Long} \geq M_{Q,Short}$ e.g. for 2003: $11 \geq 2$ months).

(a) Initiation time

	Variable (v)	NUTS-1 I_V (months after start episode)	Local $I_{V,Quick}$ (months after start episode)	Local $I_{V,Slow}$ (months after start episode)
2003	P1	2	2	3
	SM	3	2	4
	Q	6	2	12
	GW	10	4	-
2015	P1	7	2	7
	SM	7	2	7
	Q	7	2	9
	GW	11	6	-
2018	P1	4	2	6
	SM	4	3	7
	Q	5	3	10
	GW	10	3	20

(b) Maximum duration

	Variable (v)	NUTS-1 M_V (Months)	Local $M_{V,Short}$ (Months)	Local $M_{V,Long}$ (Months)
2003	P1	2	2	4
	SM	7	2	11
	Q	4	2	9
	GW	7	0	20
2015	P1	1	1	4
	SM	6	2	8
	Q	6	1	8
	GW	3	0	7
2018	P1	6	2	6
	SM	9	3	10
	Q	7	2	10
	GW	8	1	16

Legend

Propagation (P1 → SM → Q → GW)

In line with hypothesis (e.g. $I_{P1} \leq I_{SM}$; $M_{P1} \leq M_{SM}$)

Not in line with hypothesis (e.g. $I_{SM} \leq I_Q$; $M_{SM} \leq M_Q$)

In line with hypothesis (e.g. $I_Q \leq I_{GW}$; $M_Q \leq M_{GW}$)

In line with hypothesis (e.g. $I_{SM,Slow} \geq I_{Q,Quick}$)

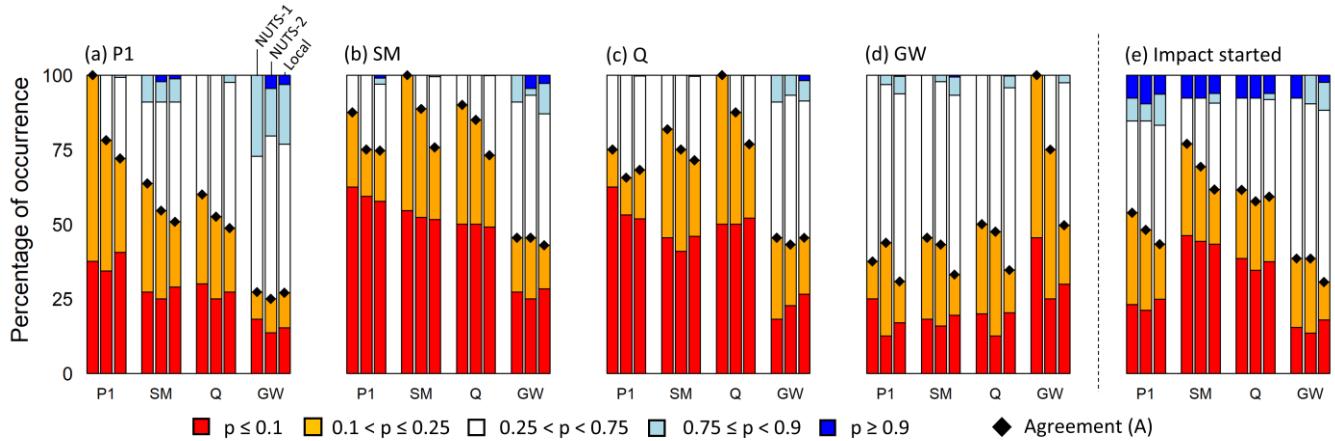
Figure 6. Typical NUTS-1 average and local drought propagation characteristics (a) initiation time (I_V, $I_{V,Quick}$ & $I_{V,Slow}$; in months after the start of the indicated year), and (b) maximum duration (M_V, $M_{V,Short}$ & $M_{V,Long}$; months) derived from ~~the~~ NUTS-1 average and local percentile time series of monthly precipitation (P1) simulated soil moisture (SM) streamflow (Q) and groundwater (GW) for the three prominent drought episodes. Colors reflect whether drought propagation characteristics are in line with our hypothesis (Sect. 2.5).

3.3 Agreement among drought information sources: the case of the drought of 2018-2019

The variability in propagating drought signals affects the agreement between different drought information sources. This agreement (*A*) among different drought information sources is indicative of the unique value of an individual source as indicator of drought occurrence. In general, this agreement as well as the classified percentiles of different variables at different scales, reveal that hydrometeorological conditions can be quite variable whenever NUTS-1 average conditions of one variable indicated drought (Fig. 7). Strongest agreement was found between NUTS-1 and NUTS-2 or local percentile time series of the same variable for P1, SM and Q (Fig. 7a-c). Whenever the NUTS-1 average percentile time series of P1, SM or Q were in drought, NUTS-2 average and local percentile time series of these variables often showed drought conditions as well (*A* = 75-

90%). However, lesser agreement was observed between NUTS-1 average GW and NUTS-2 average and local GW ($A = 50$-

395 70%; Fig. 7d). Lesser agreement was also observed among NUTS-1 average and NUTS-2 average or local percentile time series of different variables. Whenever NUTS-1 average P1 was in drought, NUTS-1 average, NUTS-2 average and local percentile time series of other variables were as well for 25-60% of cases (Fig. 7a). SM and Q agree relatively well with each other ($A = 70$-85%) but not so much with P1 and GW (A= 50-80%; Fig. 7b-c). NUTS-1 average GW often did not agree with percentile time series of other variables ($A = 30$-50%; Fig. 7d). The agreement between impact start and drought hazard

400 information reflects how well different drought hazard information sources can predict the start date of an impact. In general, it can be seen that hydrometeorological conditions can be quite variable at the start date of an impact (Fig. 7e). Strongest agreement between impact start and drought hazard occurrence was found for SM and Q ($A = 60$-75%). Lesser agreement was found between impact start and drought hazard occurrence according to P and GW ($A = 30$-55%).

405 **Figure 7. Classified percentile (*p*) distributions of different variables (x-axis) at different scales (bar triplets: NUTS-1, NUTS-2 and local) during the drought episode of 2018-19 for all months with NUTS-1 average drought conditions (*p*v<0.25) according to (a) monthly precipitation (P1), (b) simulated soil moisture (SM), (c) streamflow (Q), (d) groundwater (GW), or (e) all months in which at least one impact started.**

The medium agreement between various drought hazard information sources (Fig. 7a-d) and the fact that not one single drought

410 information source fully agreed with the start of drought impacts (Fig. 7e) motivates a further exploration of the advantages of using local multivariate over single variable regional average drought information (Fig. 8). Characterizing the drought of 2018-2019 by single variable NUTS-1 average drought information provides an incomplete picture and fails to predict part of the drought impacts (Fig. 8a,c). For example, NUTS-1 average P1 picked-up the peak of the drought in 2018 (Fig. 8a). However, NUTS-1 average P1 missed the prolonged duration of the NUTS-1 average GW drought in 2019 and associated impacts, i.e.,

415 NUTS-1 average P1 does indicate drought recovery in winter 2018-2019. In contrast, NUTS-1 average GW missed the occurrence of drought in the summer of 2018 and fails to predict the manifold of impacts that occurred. NUTS-1 average P1 also missed the occurrence of some NUTS-1 average SM and Q drought months and associated impacts in the summer of 2019. The difference between NUTS-1 average and local drought information revealed that average drought information

missed the early onset and delayed recovery of part of the local hydrological drought conditions. According to the regional average drought signal, the drought of 2018 started in April-May (Fig. 8a). However, local drought information revealed that drought conditions developed earlier for part of the study region, which matches with the earlier start of impacts related to public water supply in winter as reported for some NUTS-2 regions (Fig. 8b,c). The same was observed for GW, i.e., the NUTS-1 average GW drought of 2018 started in autumn, whereas local GW conditions reached drought much earlier, which matches with the earlier occurrence of local summer impacts in 2018 that described wells running dry.

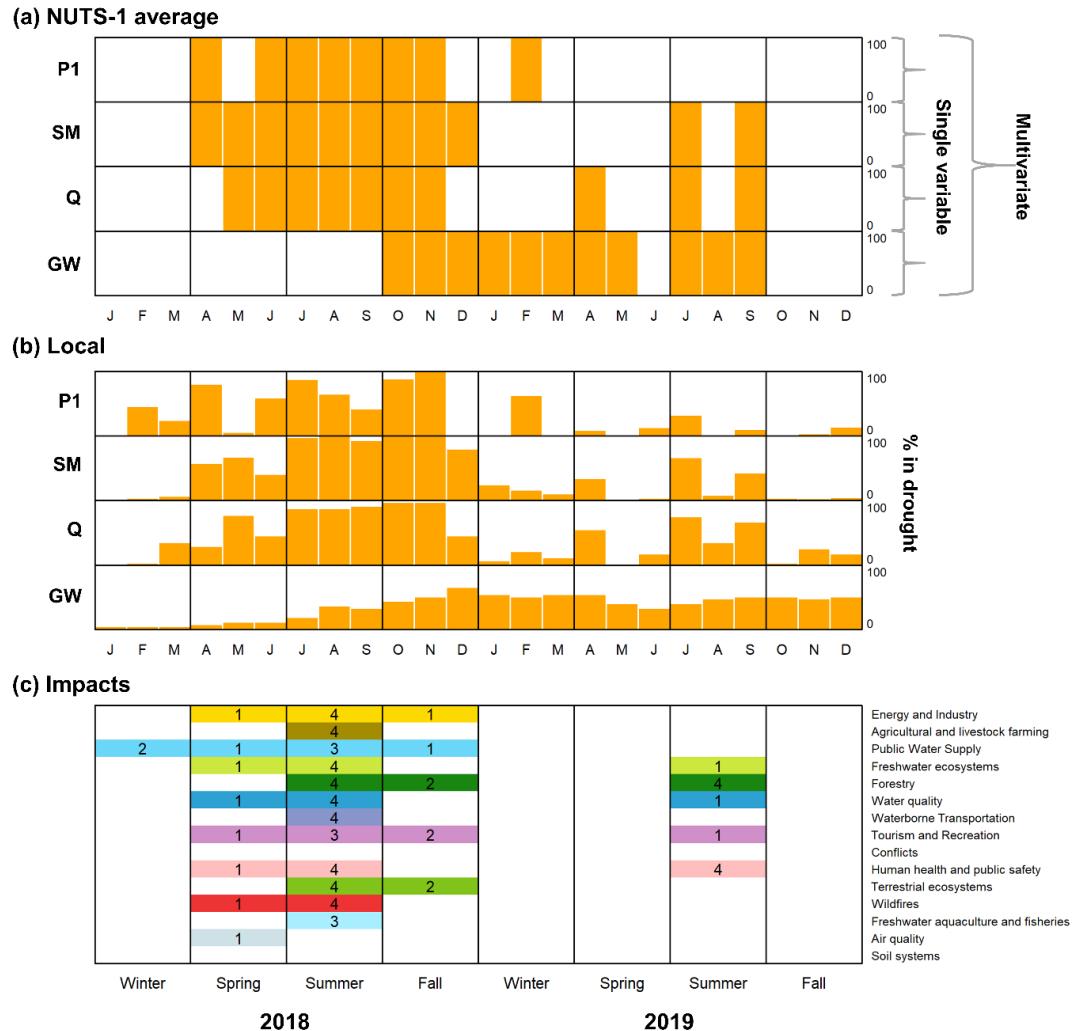



Figure 8. Differences among drought information sources displayed for the drought of 2018-2019 ranging from regional average single variable information to local multivariate drought information of precipitation (P1), simulated soil moisture (SM), streamflow (Q) and groundwater (GW): (a) NUTS-1 average and (b) local monthly drought hazard information expressed as the percentage of grid cells, catchments, or wells in drought. (c) Start of drought impacts by impact category and season, with numbers indicating for how many NUTS-2 regions impacts were reported (maximum 4).

4 Discussion

4.1 Typical droughts and their hazard and impact characteristics

Our first objective was to identify and characterize past drought episodes in Baden-Württemberg based on a multi-variable assessment. The analyses revealed different types of drought, of which the intense multi-seasonal dry spells peaking in summer (Type I) that occurred in 2003, 2015 and 2018 were most prominent according to the duration and severity of the drought hazard and the amount and variety in drought related impacts (Fig. 2-4). The episodes of 2003 and 2015 are relatively well known and documented (e.g., Ionita et al., 2017; Laaha et al., 2017; Van Lanen et al., 2016). Compared to these episodes, the drought of 2018 in the study area was generally more extreme in terms of the duration and severity of precipitation, soil moisture and river flow deficits, which is in line with findings for Northern Europe and Switzerland (resp. ectively Bakke et al., 2020; Brunner et al., 2019). Our study further revealed a multi-year nature of the drought of 2018 that made the event more impactful, which is in line with drought impacts that were reported in the media for Germany over this prolonged period (De Brito et al., 2020), and findings for the groundwater drought in the Netherlands (Brakkee et al., 2021). Hydrological droughts for part of the catchments and wells persisted far into 2019. This delayed development and prolonged recovery of hydrological drought has been reported for previous drought episodes in various studies (e.g., Parry et al., 2016; Peters et al., 2005).

The Type I drought episodes of 2003, 2015 and 2018 all coincided with some months in the growing season with above normal air temperatures. These above normal air temperatures revealed to have a compounding effect on drought impacts, contributed to soil moisture- or hydrological-drought development (e.g., Brunner et al., 2021) and could have caused a more rapid intensification of drought conditions also known as "flash droughts" (e.g., Nguyen et al., 2019). The generally increasing air temperatures will exacerbate drought and its impacts and present a challenge for future drought management (Brunner et al., 2021; Markonis et al., 2021; Pendergrass et al., 2020). The type I events studied here might be seen as a precursor of intense warm-climate drought events and their typical impacts.

The impacts of Type I episodes in the study area were severe and diverse and thanks to the data collection in impact databases such as the EDII well documented. Based on the textual descriptions provided with each impact, we discuss the type of impacts that occurred and their causes. Reported impacts in the sector "agriculture and livestock farming" were related to e.g., reduced crop yield and quality, a lack of food for livestock, or increased cost for irrigation water. Losses in agriculture are mostly attributed to low soil water availability and heat (see also e.g. Peichl et al., 2019), whereas the increased cost in irrigation relate to low (sub-)surface water levels and the consequent need to use alternative water sources. Impacts on forestry relate to tree growth and vitality and were associated with low soil moisture and groundwater levels in the impact reports, depending on the kind and age of the tree (e.g. Skiadaresis et al., 2019). Energy and industry impacts relate to reduced hydropower production due to low surface water availability or the shutdown of hydropower plants for technical or ecological reasons, e.g., a lack of cooling water or the exceedance of ecological thresholds (see also e.g. Van Vliet et al., 2016). Furthermore, streams with a number of hydropower plants experienced hydropeaking and emerging water conflicts among users with different interests, e.g., energy production versus ecology (see also e.g. Bruder et al., 2016). Public water supply presented challenges during

drought, e.g. resulting in the use of alternative resources such as other wells or rivers. Nevertheless, water security was strong
465 thanks to regional water exchange networks and long-distance water transfers (Blauhut et al., 2020). Merely few high-altitude settlements had to be supported with deliveries. Impacts on aquatic and terrestrial ecosystems and ecology were also diverse, ranging from fish die-off to the spreading of plant diseases favored by drought conditions. We found that the drivers of these impacts were above normal temperatures (heat stress) and below normal water availability. Overall, the manifold of causes of a large variety of drought impacts highlights the need for multivariate drought management.

470 ~~Lesser known and studied are the other drought types.~~ Drought types other than Type I droughts often receive less attention by media or in research. However, according to the EDII, these drought types can also be associated with a variety of impacts and, therefore, deserve more attention and awareness. Type I→II droughts are mainly characterized by prolonged drought conditions in less responsive hydrological systems that develop after intense Type I droughts (Fig. 3, 5). Notable is their asynchronous development and recovery, i.e., hydrological drought can develop and recover long after meteorological and soil
475 moisture drought. This behavior might not be accurately represented in hydrological models, particularly some of the large-scale ones (Tallaksen and Stahl 2014). This asynchronous development meant that the slowly developing hydrological drought did not necessarily coincide with droughts in other domains of the hydrological cycle or with heatwaves. This meant that the impacts of Type I→II droughts were different compared to e.g. Type I droughts (Fig. 4). First, the number of reported impacts and categories is lower suggesting that Type I→II are less impactful and visible. Second, different impact categories comprise
480 larger shares in the overall impact category distribution. For example, impacts starting in the year 2019 mostly relate to forestry, as prolonged hydrological (groundwater) drought episodes might be especially critical for some tree species (Tegel et al. 2020). Furthermore, the effects of drought on trees can be creeping and accumulating; drought weakened trees can be affected by a variety of pests and diseases which can result in delayed diebacks (Schuldt et al. 2020). We further expect hydrologically induced impacts on e.g., public water supply that started during the Type I droughts to continue. However, this needs to be
485 further investigated, as most impacts only report a start and no end date, which aligns with the general challenge of identifying when a drought and its impacts are fully recovered (e.g. Parry et al., 2016). In addition, the survey to gather additional impacts on the public water supply sector for the impact database ended in 2018. Altogether, a relatively wet winter after a Type I drought episode might give a false sign of drought recovery, as some less visible hydrological deficits and consequent impacts linger on.

490 The Type II groundwater droughts of the early 1990s exceeds other groundwater droughts in both duration and severity (Fig. 3). These prolonged groundwater drought conditions might be associated with the long-term meteorological water deficits and the absence of distinct wet periods (Fig. 2), as also reported for parts of central Europe (e.g. Hannaford et al., 2011; Spinoni et al., 2015). Nevertheless, (changes in) groundwater abstraction cannot be ruled out for all wells and might have had an influence. Groundwater related impacts for this period were not available in the EDII for Baden-Württemberg. This absence of impact
495 reports in the EDII mostly relates to the lower research attention towards droughts in this period as well as to the lower impact data availability prior to the digital era. Nevertheless, neighboring NUTS-1 regions show evidence of impacts of this episode in the EDII related to e.g., cargo transport on the river Rhine or tree vitality.

Type III droughts describe shorter meteorological dry spells mainly affecting soil moisture and streamflow (Fig. 3). These meteorological dry spells were not intense enough to cause a strong decline in groundwater but could nevertheless coincide with and worsen ongoing groundwater droughts (Type III+II). The Type III episode of 2011 and 2014 occurred before the summer season. Their timing in occurrence, together with their relatively short duration, affected the impact categories that occurred, which were mostly related to energy (hydropower) and water use conflicts. Typical summer impacts enhanced by high air temperatures related to e.g., water quality or agriculture were absent. Remarkable about the drought of 2011 was the large share of impacts related to waterborne transportation, which related to low water levels in the Rhine towards the end of spring but also in a very dry November (see also Kohn et al., 2014). Type III droughts also occurred in the winters of for example 2005-2006 and 2016-2017. We did not consider these winter droughts in the results section, because of their limited range in impacts due to their timing. The Type III+II episode of 1998 revealed the joint occurrence of soil moisture, streamflow, and groundwater drought. There is not much evidence in the EDII on the impacts of this episode given the general lower impact availability prior to the year 2000. Nevertheless, we hypothesize that such Type III+II episodes provide extra stress for e.g., irrigated agriculture or tree vitality, given the joint occurrence of limited soil moisture, streamflow, and groundwater supply.

Overall, different drought types share common characteristics and impacts. This similarity among drought types highlights the potential of drought type specific management and research (Markonis et al., 2021). Our results imply that such a typology should not only consider the timing of drought, but also the domain of the hydrological cycle in which the drought appeared, and whether droughts in different domains of the hydrological cycle as well as high temperatures coincided or not. Most well-known and impactful were the Type I drought episodes, and these episodes can be adopted as worst-case benchmarks for drought management. However, the Type I drought episodes of the past decades never coincided with persistent long-term hydrological (groundwater) drought. An important question remains what would have happened when a drought episode like 2018 started when storages in hydrological systems were low, i.e., a Type I+II episode. Stress tests or scenario studies can be used to explore the impacts of such worst-case episodes as well as to prepare drought management for the future (e.g Grecksch, 2019; Hellwig et al., 2021; Stoelzle et al., 2020).

4.2 Regional drought propagation characteristics

Our second aim was to evaluate the (variability in) regional drought propagation signals. We showed that drought propagation of Type I droughts of 2003, 2015 and 2018 generally followed the hypothetical order and time lag of drought propagation as described in e.g., Van Loon (2015). This suggests that the concept of drought propagation could be a general guiding principle for regional drought management. However, we also found several deviations from the general concept of drought propagation that need to be considered.

First, drought propagation does not have to complete the full cycle, which means that the expected ordering and lengthening is not always observed (Fig. 2, 5-6). For example, the relatively short meteorological droughts of 2011 and 2014 propagated to below normal soil moisture and river flow but groundwater levels often stayed in the normal range, especially for the less

responsive aquifer types. On the other hand, short dry spells can severely threaten groundwater systems when the initial conditions are low at the start of the dry spell as occurred in, e.g., summer 1998. This implies the importance of tracking non-drought conditions (approaching drought) in a drought management context, especially for the more slowly responding rivers and aquifers. For these slowly responding hydrological systems, the initial conditions can be an important precursor for the likelihood of future drought conditions (e.g. Parry et al., 2018). Another deviation from the general drought propagation concept is the occasional absence of lengthening in propagating drought signals, which relates to the time-lag in the development of deficits and consequent chances of having drought-recovering wet conditions prior to drought reaching its full extent. For example, we showed for the prominent drought episode of 2015 that groundwater reached drought conditions after a prolonged meteorological dry spell and associated persistent soil moisture and river flow droughts (Fig. 5-6). However, a very wet period followed soon after the development of the still relatively small groundwater deficits, which meant a quick recovery. Finally, the spatial variation in drought conditions revealed that local drought propagation over a larger region does not necessarily follow the expected sequencing of different drought types (Fig. 6). For example, a responsive groundwater aquifer with a low amount of storage reached below normal conditions before a less responsive river underlain by a high storage aquifer (see also Stoelzle et al., 2014). The spatially variable response time was also visible in the impacts, e.g., the start dates of local impacts on public water supply in 2018 ranged from early to late in the year (Fig. 8c). These variable hydrological responses to meteorological dry spells underpin the need for the consideration of local differences in response to meteorological drought among different (hydrological) systems at different timescales (Vicente-Serrano et al. 2021; Wu et al., 2021).

4.3 Disagreement among drought information sources

Our third aim was to assess the variability in drought hazard signals and consequent disagreement among drought information sources. The found variability in propagating drought signals across variables and scales implies complexities in the use of single variable or regional average (composite) drought information for comprehensive drought assessments, as relevant variations are lost (Fig. 7-8). Single variable or regional average drought hazard information, especially of soil moisture and streamflow, could predict the start of part of the drought impacts (Fig. 7e). However, regional average single variable drought hazard information was not sufficient to predict the start of all drought impact occurrences. Part of the reason in the mismatch between drought hazard information and the start of drought impacts might be related to inaccuracies in reporting. However, another explanation might be the mismatch between drought hazard information sources, e.g., regional single variable information might miss the earlier onset or delayed recovery of some (hydrological) systems- (Fig. 8). This suggests, together with the variety of causes of drought impacts (Sect. 4.1), that the joint consideration of different variables at a local rather than regional scale is needed to predict the full range of impacts during different stages of drought. Drought monitoring and management would benefit from future efforts towards joint portals that include near real-time multivariate drought hazard and impact information at different scales. Such portals might be used to design or implement operational definitions of drought, paving the way towards more targeted, sector and location specific, drought management.

5 Conclusion

565 Past drought episodes and their impacts in southern Germany ranged from intense multi-seasonal drought episodes with severe soil moisture and streamflow deficits and compounding heat waves to less intense but more prolonged multi-year drought episodes and associated low groundwater levels. The identified drought types each share common hazard, propagation, and impact characteristics, which suggests that drought-type specific management options could be designed. Of the different drought types, intense multi-seasonal drought episodes with compounding heat waves (as in 2003, 2015 and 2018) caused the

570 largest number and variety of impacts. This drought type might become a challenge for future drought management in a warming climate. The drought of 2018, as an example, stood out by the length of the meteorological dry spell and consequent prolonged and severe soil moisture and river flow deficits. Groundwater deficits in 2018 generally developed slower but lasted well beyond the end of 2018. This multi-year nature of drought highlights the need of ~~long-term drought management~~continuous drought monitoring beyond the peak of the drought in summer.

575 Drought ~~risk management components such as~~ monitoring and early warning is an important component of risk management, but often ~~rely~~ relies on assumptions of drought propagation ~~assumptions~~ and on concepts such as sequencing, time lag and lengthening. Studying the regional propagation of droughts mostly confirmed the ~~existence~~ validity of these concepts also on a regional scale and can thereby be expected to be a useful general guide to raise regional drought ~~management~~awareness for multiple water use sectors. However, we also found deviations from the general drought propagation concept as drought

580 propagation did not always affect the full hydrological cycle and the length of the dry spell did not always increase. Further, the order of appearance and lengthening was not necessarily visible in local drought conditions e.g., streamflow drought in a less responsive river might develop later and persist longer compared to groundwater drought in a more responsive aquifer. Regional average single variable drought hazard information favors easy interpretation. However, drought information derived from different variables at different scales can show a low agreement. Therefore, in any region as diverse as the case study

585 shown here a drought assessment based on only a regional average ~~single variable~~ drought information ~~can vary significantly from a multivariate assessment of local drought conditions and might consequently~~may ~~fail~~not be sufficient to explain ~~predict~~ the occurrence of some of the impacts. ~~Overall, comprehensive drought management and assessments benefit from the appraisal of local multivariate drought signals.~~

590 **Competing interests.** The authors declare no competing interests.

Data ~~availabillity~~availability. ~~Data are available from sources mentioned in Section 2.2 or upon request from the providers cited in this Section.~~ Local percentile timeseries of all drought hazard variables used in this manuscript (P1, P3, P12, SM, Q, GW) can be derived from 10.5281/zenodo.6449934.

595

Acknowledgements. This work contributes to the DRIeR project supported by the Wassernetzwerk Baden-Württemberg (Water Research Network), which is funded by the Ministerium für Wissenschaft, Forschung und Kunst Baden-Württemberg (Ministry of Science, Research, and the Arts of the State Baden-Württemberg) (grant no. AZ. 7532.21/2.1.6). We further acknowledge the DWD and LUBW for providing data. Financial support of the German Research Foundation (DFG) for SDS@hd - Scientific Data Storage is acknowledged. All analyses were carried out with the open-source software R (https://www.r-project.org/).

Author contributions. Conceptualization (all), formal analyses and writing of first draft (ET), review and editing (VB, MS, LM & KS), visualization (ET, VB, MS), funding acquisition (LM, KS).

References

Bakke, S. J., Ionita, M., Tallaksen, L. M.: The 2018 Northern European Hydrological Drought and Its Drivers in a Historical Perspective. Hydrol. Earth Syst. Sci., 24, 5621–5653, https://doi.org/10.5194/hess-24-5621-2020, 2020.

Barker, L. J., Hannaford, J., Chiverton, A., Svensson, C.: From Meteorological to Hydrological Drought Using Standardised Indicators. Hydrol. Earth Syst. Sci., 20, 2483–2505, https://doi.org/10.5194/hess-20-2483-2016, 2016.

Blauhut, V., Stahl, K., Falasca, G.: Dürre Und Die Öffentliche Wasser- Versorgung in Baden-Württemberg: Folgen, Umgang Und Wahrnehmung. Wasserwirtschaft, 11, 31-36, 2020.

Blauhut, V., Gudmundsson, L., Stahl, K.: Towards Pan-European Drought Risk Maps: Quantifying the Link between Drought Indices and Reported Drought Impacts. Environ. Res. Lett., 10, 014008, https://doi:10.1088/1748-9326/10/1/014008, 2015.

Bloomfield, J. P. Marchant, B. P.: Analysis of groundwater drought building on the standardised precipitation index approach. Hydrol. Earth Syst. Sci., 17, 4769–4787, https://doi.org/10.5194/hess-17-4769-2013, 2013.

Brakkee, E., van Huijgevoort, M., and Bartholomeus, R. P.: Spatiotemporal development of the 2018–2019 groundwater drought in the Netherlands: a data-based approach, Hydrol. Earth Syst. Sci. Disc., https://doi.org/10.5194/hess-2021-64, 2021.

Bruder, A, Tonolla, D., Schweizer, S. P., Vollenweider, S., Langhans, S. D., Wüest, A.: A Conceptual Framework for Hydropeaking Mitigation. Sci. Tot. Environ., 568, 1204-1212, https://doi.org/10.1016/j.scitotenv.2016.05.032, 2016.

Brunner, M. I., Liechti, K., Zappa, M.: Extremeness of Recent Drought Events in Switzerland: Dependence on Variable and Return Period Choice. Nat. Hazard Earth Sys. Sci., 19, 2311–2323, https://doi.org/10.5194/nhess-19-2311-2019, 2019.

Brunner, M. I., Swain, D. L., Gilleland, E., Wood, A. W.: Increasing Importance of Temperature as a Contributor to the Spatial Extent of Streamflow Drought. Environ. Res. Lett., 16, 024038, https://10.1088/1748-9326/abd2f0, 2021.

Changnon, S. A.: Detecting Drought Conditions in Illinois. Illinois State Water Survey Champaign, Illinois, US, 1987.

De Brito, M. M., Kuhlicke, C., & Marx, A: Near-real-time drought impact assessment: a text mining approach on the 2018/19 drought in Germany. Environmental Research Letters, 15, 1040a9, https://doi.org/10.1088/1748-9326/aba4ca, 2020.

DWD: Climate and Environment Observations, http://opendata.dwd.de/climate_environment/CDC/, last access: 04.03.2021, 2020.

Erfurt, M., Glaser, R., Blauhut, V.: Changing impacts and societal responses to drought in southwestern Germany since 1800.
630 Reg. Environ. Change, 19, 2311-2323, https://doi.org/10.1007/s10113-019-01522-7, 2019.

Erfurt, M., Skiadaresis, G., Tijdeman, E., Blauhut, V., Bauhus, J., Glaser, R., Schwarz, J., Tegel, W., and Stahl, K.: A multidisciplinary drought catalogue for southwestern Germany dating back to 1801, Nat. Hazards Earth Syst. Sci., 20, 2979–2995, https://doi.org/10.5194/nhess-20-2979-2020, 2020.

Grecksch, K.: Scenarios for Resilient Drought and Water Scarcity Management in England and Wales. International Journal
635 of River Basin Management, 17, 219-217, https://doi.org/10.1080/15715124.2018.1461106, 2019.

Hannaford, J., Lloyd-Hughes, B., Keef, C., Parry, S., Prudhomme, C.: Examining the Large-Scale Spatial Coherence of European Drought Using Regional Indicators of Precipitation and Streamflow Deficit. Hydrol. Process., 25, 1146-1162, 10.1 https://doi.org/002/hyp.7725, 2011.

Hao, Z., Singh, V. P.: Drought Characterization from a Multivariate Perspective: A Review. J. Hydrol., 527, 668-678,
640 https://doi.org/10.1016/j.jhydrol.2015.05.031, 2015.

Hari, V., Rakovec, O., Markonis, Y., Hanel, M., Kumar, R.: Increased Future Occurrences of the Exceptional 2018–2019 Central European Drought under Global Warming. Sci. Rep., 10, 12207, https://doi.org/10.1038/s41598-020-68872-9, 2020.

Hellwig, J., Stoelzle, M., Stahl, K.: Groundwater and baseflow drought responses to synthetic recharge stress tests, Hydrol. Earth Syst. Sci., 1053–1068, https://doi.org/10.5194/hess-25-1053-2021, 2021.

645 Heudorfer, B., Haaf, E., Stahl, K., Barthel, R.: Index-Based Characterization and Quantification of Groundwater Dynamics. Water Resour. Res., 55, 5575-5592, https://doi.org/10.1029/2018WR024418, 2019.

Hisdal H, Tallaksen L. Drought event definition. Technical Report, ARIDE Technical Report No. 6, University of Oslo, Norway, 2000.

Ionita, M., Tallaksen, L. M. Kingston, D. G., Stagge, J. H., Laaha, G., Van Lanen, H. A. J., Scholz, P., Chelcea, S. M.,Haslinger,
650 K.: The European 2015 Drought from a Climatological Perspective. Hydrol. Earth Syst. Sci., 21, 1397–1419, https://doi.org/10.5194/hess-21-1397-2017, 2017.

Kendon, M., Marsh, T., Parry, S.; The 2010-2012 Drought in England and Wales. Weather, 68, 88–95, https://doi.org/10.1002/wea.2101, 2013.

Knutson, C., Hayes, M., Phillips, T.: How to Reduce Drought Risk, Western Drought Coordination Council, Preparedness and
655 Mitigation Working Group, National Drought Mitigation Center, Nebraska, USA, 1998.

Kohn, I., Rosin, K., Freudiger, D., BeIz, J. U., Stahl, K., Weiler, M.: Niedrigwasser in Deutschland 2011. Hydrologie Und Wasserbewirtschaftung, 58, 4-17, https://doi.org/10.5675/HyWa_2014,1_1, 2014.

Laaha, G., Gauster T., Tallaksen, L. M. Vidal, J. P., Stahl, K., Prudhomme, C., Heudorfer, B., Vlnas, R., Ionita, M., Van Lanen, H. A. J., Adler, M. J., Caillouet, L., Delus, C., Fendekova, M., Gailliez, S., Hannaford, J., Kingston, D. Van Loon, A. F.

660   Mediero, L., Osuch, M., Romanowicz, R., Sauquet, E., Stagge, J. H. and Wong, W. K.: The European 2015 Drought from a Hydrological Perspective. Hydrol. Earth Syst. Sci., 21, 3001–3024, https://doi.org/10.5194/hess-21-3001-2017, 2017.

Lloyd-Hughes, B.: The Impracticality of a Universal Drought Definition. Theor. Appl. Climatol., 117, 607–611, https://doi.org/10.1007/s00704-013-1025-7, 2014.

LUBW: Bewertungsmessnetz Grundwasservorräte. https://guq.lubw.baden-wuerttemberg.de/, last access: 04.03.2021, 2020.

665   Markonis, Y., Kumar, R., Hanel, M., Rakovec, O., Máca, P., AghaKouchak, A.: The Rise of Compound Warm-Season Droughts in Europe. Sci. Adv., 7, eabb9668, https://10.1126/sciadv.abb9668, 2021.

Nguyen, H., Wheeler, M. C., Otkin, J. A., Cowan, T., Frost, A., & Stone, R.: Using the evaporative stress index to monitor flash drought in Australia. Environ. Res. Lett., 14, 064016, https://doi.org/10.1088/1748-9326/ab2103, 2019.

Parry, S., Wilby, R. L., Prudhomme, C., Wood, P. J.: A Systematic Assessment of Drought Termination in the United

670   Kingdom. Hydrol. Earth Syst. Sci., 20, 4265–4281, https://doi.org/10.5194/hess-20-4265-2016, 2016.

Parry, S., Wilby, R. L., Prudhomme, C., Wood, P., McKenzie, A.: Demonstrating the Utility of a Drought Termination Framework: Prospects for Groundwater Level Recovery in England and Wales in 2018 or Beyond. Environ. Res. Lett., 13, 064040, https://doi.org/10.1088/1748-9326/aac78c, 2018.

Peichl, M., Thober, S. Samaniego, L., Hansjürgens, B., Marx, A.: Climate Impacts on Long-Term Silage Maize Yield in

675   Germany. Sci. Rep., 9, 7674, https://doi.org/s41598-019-44126-1, 2019.

Pendergrass, A. G., Meehl, G. A., Pulwarty, R., Hobbins, M., Hoell, A., AghaKouchak, A., …, and Woodhouse, C. A.: Flash droughts present a new challenge for subseasonal-to-seasonal prediction. Nat. Clim. Change, 10, 191-199, 10.1038/s41558-020-0709-0, 2020.

Peters E. Propagation of drought through groundwater systems: illustrated in the Pang (UK) and Upper-Guadiana (ES)

680   catchments. PhD Thesis, Wageningen University, the Netherlands, 2003.

Peters, E., Van Lanen, H. A. J. , Torfs, P. J. J. F., Bier., G.: Drought in Groundwater - Drought Distribution and Performance Indicators. J. Hydrol., 306, 302–317, https://doi.org/10.1016/j.jhydrol.2004.09.014, 2005.

Rauthe, M., Steiner H., Riediger, U., Mazurkiewicz, A., Gratzki, A.: A Central European Precipitation Climatology - Part I: Generation and Validation of a High-Resolution Gridded Daily Data Set (HYRAS). Meteorol. Z., 22, 235-256,

685   https://doi.org/10.1127/0941-2948/2013/0436, 2013.

Samaniego, L., Thober, S., Kumar, R., Wanders, N., Rakovec, O., Pan, M., Zink, M., Sheffield, J., Wood, E. F., Marx, A.: Anthropogenic Warming Exacerbates European Soil Moisture Droughts. Nat. Clim. Change, 8, 421–426, https://doi.org/10.1038/s41558-018-0138-5, 2018.

Schuldt, B., Buras, A., Arend, M., Vitasse, Y., Beierkuhnlein, C., Damm, A., Gharun, M., Grams, T. E. E., Hauck, M., Hajek,

690   P., Hartmann, H., Hiltbrunner, E., Hoch, G., Holloway-Phillips, G., Körner, C., Larysch, E., Lübbe, T., Nelson, D. B., Rammig, A., Rigling, A., Rose, L., Ruehr, N. K., Schumann, K., Weiser, F., Werner, C., Wohlgemuth, T., Zang, C. S., Kahmen, A.: A First Assessment of the Impact of the Extreme 2018 Summer Drought on Central European Forests. Basic Appl. Ecol., 45, 86-103, https://doi.org/10.1016/j.baae.2020.04.003, 2020.

Schwarz, J. A., Skiadaresis, G., Kohler, M., Kunz, J., Schnabel, F., Vitali, V., Bauhus, J.: Quantifying Growth Responses of Trees to Drought—a Critique of Commonly Used Resilience Indices and Recommendations for Future Studies. Current Forestry Reports, 6, 185-200, https://doi.org/10.1007/s40725-020-00119-2, 2020.

695

Skiadaresis, G., Schwarz, J. A., Bauhus, J.: Groundwater Extraction in Floodplain Forests Reduces Radial Growth and Increases Summer Drought Sensitivity of Pedunculate Oak Trees (Quercus Robur L.). Frontiers in Forests and Global Change, https://doi.org/10.3389/ffgc.2019.00005, 2019.

700 Spinoni, J., Naumann, G., Vogt, J. V., Barbosa, P.: The Biggest Drought Events in Europe from 1950 to 2012. J. Hydrol.: Regional Studie, 3, 509-524, https://doi.org/10.1016/j.ejrh.2015.01.001, 2015.

Stahl, K., Kohn, I., Blauhut, V., Urquijo, J., De Stefano, L., Acácio, V., Dias, S., Stagge, J. H., Tallaksen, L. M., Kampragou, E., Van Loon, A. F., Barker, L. J., Melsen, L. A., Bifulco, C., Musolino, D., de Carli, A., Massarutto, A., Assimacopoulos, D., and Van Lanen, H. A. J.: Impacts of European drought events: insights from an international database of text-based reports.

705 Nat. Hazards Earth Syst. Sci., 16, 801–819, https://doi.org/10.5194/nhess-16-801-2016, 2016.

Stephan, R., Erfurt, M., Terzi, S., Žun, M., Kristan, B., Haslinger, K., and Stahl, K.: An Alpine Drought Impact Inventory to explore past droughts in a mountain region, Nat. Hazards Earth Syst. Sci. Discuss., https://doi.org/10.5194/nhess-2021-24, 2021.

Stoelzle, M., Stahl, K., Morhard, A., Weiler, M.: Streamflow sensitivity to drought scenarios in catchments with different

710 geology. Geophys. Res. Lett., 41, 6174-6183, https://doi.org/10.1002/2014GL061344, 2014.

Stoelzle, M., Staudinger, M., Stahl, K., and Weiler, M.: Stress testing as complement to climate scenarios: recharge scenarios to quantify streamflow drought sensitivity, Proc. IAHS, 383, 43–50, https://doi.org/10.5194/piahs-383-43-2020, 2020.

Stoelzle, M., Weiler, M., Stahl, K., Morhard, A., Schuetz, T.: Is There a Superior Conceptual Groundwater Model Structure for Baseflow Simulation? Hydrol. Process., 29, 1301-1313, https://doi.org/10.1002/hyp.10251, 2015.

715 Tallaksen, L. M., Van Lanen, H. A. J.: Hydrological Drought, Processes and Estimation Methods for Streamflow and Groundwater. In: Developments in Water Science, vol. 48., Elsevier Science B.V., Amsterdam, the Netherlands, 2004.

Tallaksen, L. M., Stahl, K.: Spatial and Temporal Patterns of Large-Scale Droughts in Europe: Model Dispersion and Performance. Geophys. Res. Lett., 41, 429–434, https://doi.org/10.1002/2013GL058573, 2014.

Tegel, W., Seim, A., Skiadaresis, G., Ljungqvist. F. C., Kahle, H. P., Land, A., Muigg, B., Nicolussi, K., Büntgen, U.: Higher

720 Groundwater Levels in Western Europe Characterize Warm Periods in the Common Era. Sci. Rep., 10, 16284, https://doi.org/10.1038/s41598-020-73383-8, 2020.

Tijdeman, E., Stahl, K. Tallaksen, L. M.: Drought Characteristics Derived Based on the Standardized Streamflow Index – a Large Sample Comparison for Parametric and Nonparametric Methods. Water Resour. Res., 56, e2019WR026315, https://doi.org/10.1029/2019WR026315, 2020.

725 Tijdeman, E. and Menzel, L.: The development and persistence of soil moisture stress during drought across southwestern Germany, Hydrol. Earth Syst. Sci., 25, 2009–2025, https://doi.org/10.5194/hess-25-2009-2021, 2021a.

Tijdeman, E., Menzel, L.: Daily Gridded Soil Moisture Simulations on a 1 Km Resolution Grid Covering Baden-Württemberg. Heidata repository, version 1, https://doi.org/10.11588/data/PRXZAS, 2021b.

730 Toreti, A., Belward, A., Perez-Dominguez, I., Naumann, G., Luterbacher, J., Cronie, O., Seguini, L., Manfron, G., Lopez-Lozano, R., Baruth, B., Van den Berg, M., Dentener. F., Ceglar, A., Chatzopoulos, T., Zampieri, M.: The Exceptional 2018 European Water Seesaw Calls for Action on Adaptation. Earth's Future, 7, 652-663, https://doi.org/10.1029/2019EF001170, 2019.

UDO: Umwelt-Daten Und -Karten Online. https://udo.lubw.baden-wuerttemberg.de/, last access: 04.03.2021, 2020.

Vicente-Serrano, S. M., Peña-Angulo, D., Murphy, C., López-Moreno, J. I., Tomas-Burguera, M., Domínguez-Castro, F.,
735 Tian, F., Eklundh, L., Cai, Z., Alvarez-Farizo, B., Noguera, I., Camarero, J. J., Sánchez-Salguero, R., Gazol, A., Grainger, S., Conradt, T., Boincean, B., El Kenawy, A.: The Complex Multi-Sectoral Impacts of Drought: Evidence from a Mountainous Basin in the Central Spanish Pyrenees. Sci. Total Environ., 769, 144702, https://doi.org/10.1016/j.scitotenv.2020.144702, 2021.

Van Lanen, H. A. J., Laaha, G., Kingston, D. G., Gauster, T., Ionita, M., Vidal, J.-P., Vlnas, R., Tallaksen, L. M., Stahl, K.,
740 Hannaford, J., Delus, C., Fendekova, M., Mediero, L., Prudhomme, C., Rets, E., Romanowicz, R. J., Gailliez, S., Wong, W. K., Adler, M.-J., Blauhut, V., Caillouet, L., Chelcea, S., Frolova, N., Gudmundsson, L., Hanel, M., Haslinger, K., Kireeva, M., Osuch, M., Sauquet, E., Stagge, J. H., and Van Loon, A. F.: Hydrology Needed to Manage Droughts: The 2015 European Case. Hydrol. Process., 30, 3097–3104, https://doi:10.1002/hyp.10838, 2016.

Van Loon, A. F.: Hydrological Drought Explained, WIRES Water, 2, 359–392, https://doi.org/10.1002/wat2.1085, 2015.

745 Van Loon, A. F., Gleeson, T., Clark, J., Van Dijk, A. I., Stahl, K., Hannaford, J., Di Baldassarre, G., Teuling, A. J., Tallaksen, L.M., Uijlenhoet, R., Hannah, D. M., Sheffield, J., Svoboda, M., Verbeiren, B., Wagener, T., Rangecroft, S., Wanders N., Van Lanen, H. A.: Drought in the Anthropocene. Nat. Geosci., 9, 89-91, https://doi.org/10.1038/ngeo2646, 2016.

Van Loon, A. F., Tijdeman, E., Wanders, N., Van Lanen, H. J., Teuling, A. J., & Uijlenhoet, R.: How climate seasonality modifies drought duration and deficit. Journal of Geophysical Research: Atmospheres, 119(8), 4640-4656,
750 https://doi.org/10.1002/2013JD020383, 2014.

Van Loon, A. F. and Van Lanen, H. A. J.: A process-based typology of hydrological drought, Hydrol. Earth Syst. Sci., 16, 1915–1946, https://doi.org/10.5194/hess-16-1915-2012, 2012.

Van Vliet, M. T. H., Sheffield, J., Wiberg, D., Wood, E. F.: Impacts of Recent Drought and Warm Years on Water Resources and Electricity Supply Worldwide. Environ. Res. Lett., 11, 124021, https://doi.org/10.1088/1748-9326/11/12/124021, 2016.

755 Wanders, N., Wada Y.: Human and Climate Impacts on the 21st Century Hydrological Drought. J. Hydrol., 526, 208–220, https://doi.org/10.1016/j.jhydrol.2014.10.047, 2015.

Weibull, W.: A Statistical Theory of Strength of Materials. Ing. Vet. Ak. Handl., Stockholm, Sweden, 1939.

Wilhite, D. A., Easterling, W. E., Wood, D. A.: Planning for Drought: Toward a Reduction of Societal Vulnerability. Routledge, New York, USA, 2019.

760 Wilhite, D. A., Glantz, M. H.: Understanding: The Drought Phenomenon: The Role of Definitions. Water Int., 10, 111–120, https://doi.org/10.1080/02508068508686328, 1985.

Wu, J., Chen, X., Yao, H., Zhang, D.: Multi-timescale assessment of propagation thresholds from meteorological to hydrological drought. Sci. Total Environ., 765, 144232, https://doi.org/10.1016/j.scitotenv.2020.144232, 2021.

Zargar, A., Sadiq, R., Naser, B., Khan, F. I.: A Review of Drought Indices. Environ. Rev., 19, 333–349, 765 https://doi.org/10.1139/a11-013, 2011.

Zscheischler, J., Martius, O., Westra, S., Bevacqua, E., Raymond, C., Horton, R. M., Van den Hurk, B., AghaKouchak, A., Jézéquel, A., Mahecha, M. D., Maraun, D., Ramos, A. M., Ridder, N. N., Thiery, W., Vignotto, E.: A Typology of Compound Weather and Climate Events. Nat. Rev. Earth & Environment, 1, 333-347, https://doi.org/10.1038/s43017-020-0060-z, 2020.