

Multiscalar and nonlinear controls of drought impacts in Spain revealed by media-reported data

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15 **Abstract.** This study assesses the multiscalar and nonlinear relationships between drought severity, characterized by the
Standardized Precipitation Index (SPI) and the Standardized Precipitation Evapotranspiration Index (SPEI), and media-
reported hydrological and agricultural impacts across Spain during 1976–2023. Drought indices were derived from a high-
resolution gridded climate dataset (1.1 km), while impact data were obtained from standardized monthly frequencies of
drought-related newspaper articles at the provincial scale. Results demonstrate robust temporal coherence between drought
20 conditions and impacts, with major drought episodes associated with anomalies exceeding +1 to +2 standard deviations.
Drought–impact relationships strengthen markedly with accumulation timescale, reaching maximum correlations at 12–~~24~~30
months ($|r| \approx 0.6$ – 0.8), while short timescales (1–3 months) show weak associations ($|r| < 0.3$). Hydrological impacts are
primarily associated with accumulated medium- and long-term moisture deficits (12–48 months), whereas agricultural impacts
are associated with shorter accumulation timescales (3–12 ~~months~~ months). However, for both sectors, dominant accumulation
25 timescales lengthen as drought severity increases. Sensitivity analyses reveal pronounced nonlinear responses, ~~with~~ the
frequency of media reports on impacts ~~media-reported impact frequency~~ increasing disproportionately during severe drought
conditions with a peak at 12–~~36~~ months. The consistently stronger association of SPEI relative to SPI, especially for
agricultural impacts, suggests a dominant contribution of atmospheric evaporative demand. ~~The consistently stronger~~
~~association of SPEI relative to SPI, especially in case of agricultural impacts, suggesting a dominant contribution of~~
30 ~~atmospheric evaporative demand.~~ Our results indicate that the frequency of media reports on impacts ~~media-reported impact~~
~~frequency~~ is associated with accumulated moisture deficits ~~across timescales~~ and temperature-dependent processes across
timescales and provide a framework for improved impact-based drought monitoring in a warming climate.

1 Introduction

35 Drought is a complex and cyclical natural hazard that can lead to severe impacts on ecosystems, water resources, agriculture
and economies (Vicente-Serrano et al., 2016; Slette et al., 2019; McGlade et al., 2019; El Kenawy et al., ~~2023~~2026). Unlike
rapidly developing hazards, drought evolves gradually over months to years and thus detection and assessment of drought
impacts ~~is~~are more difficult (Vicente-Serrano, 2016). Climate change has led to severe droughts and their consequences for
the world's population (Spinoni et al., 2021; Runde et al., 2022; Wang et al., 2023). Increased temperature and ~~increased~~
40 atmospheric evaporative demand have been reported to be linked with increased frequency, severity and duration of droughts
~~in for~~ Southern Europe (Vicente-Serrano et al., 2014; Stagge et al., 2017).

To monitor and characterize drought severity, researchers use drought indices, which provide objective measures across
meteorological, agricultural, and hydrological domains (Keyantash, 2021). The relationship between most drought ~~indexes~~
~~indices~~ and their actual impacts can be non-linear, even though they are highly correlated (Bachmair et al., 2015; Kchouk et
45 al., 2022). The magnitude and nature of drought impacts are largely influenced by local environmental and socioeconomic
conditions (Kamali et al., 2019; Wijitkosum~~-et al.~~, 2025). These impacts are diverse, including hydrological impacts such as
reduced river flows, reservoir depletion, and groundwater decline, as well as agricultural impacts such as crop failure, live stock
losses, and food insecurity (Vicente-Serrano et al., 2021; Walker et al., 2024).

Hydrological impacts are primarily controlled by accumulated precipitation deficits, driven by mechanisms including soil
50 moisture depletion, reduced groundwater recharge and reduced reservoir storage (Melo et al., 2016; Álvarez-Garretón et al.,
2021; Liu et al., 2024). Thus, hydrological impacts are largely the cumulative effects of a prolonged deficit of moisture.
Hydrological systems primarily respond to medium- and long-term drought conditions and therefore exhibit delayed yet
persistent responses (Van Loon et al., 2024). Agricultural impacts are more closely associated with short- to medium-term
droughts because soil moisture and atmospheric water demand respond rapidly during the growing season (Kimm et al., 2020;
55 Rigden et al., 2020). These sectoral differences highlight the importance of the drought accumulation timescale for the link
between climate anomalies and impacts. Warm conditions can intensify water stress through increased atmospheric
evaporative demand, leading to larger differences between precipitation-based and temperature-sensitive drought indicators
(Vicente-Serrano et al., 2014; Tomás-Burguera et al., 2020; Ji et al., 2026). Similar drought conditions do not necessarily
produce similar impacts depending on regional vulnerability, adaptive capacity and land and water management protocols
60 (Welsh et al., 2013; Engström et al., 2020; Van Schmidt et al., 2023). Together, these controls introduce nonlinear behaviour
into drought-impact relationships and may induce disproportionate effects in the event of severe or extended droughts.

Media coverage plays an important role in shaping the societal interpretation of drought impacts. Newspaper articles provide
information on the timing, location, and sectoral manifestation of drought impacts (Bachmair et al., 2016; Murphy et al., 2017;
Wang et al., 2020; O'Connor et al., 2023; Jobbová et al., 2024), which allows us to understand physical conditions and social

65 responses. They are often used for reconstruction of past droughts and interpretation of human behaviour in relation to climate change (BrázdilBrazil et al., 2018; Nash et al., 2019). Drought indices in Europe and the Mediterranean are highly correlated with reported impacts (e.g., in agriculture and water supply systems) (e.g., Stahl et al., 20152016; Blauhut et al., 2016; Shyrokaya et al., 2023). Such media reports have been a key to documenting regional drought dynamics and public perceptions in Spain (Llasat et al., 2009; Ruiz Sinoga and León Gross, 2013; Serrano-Acebedo et al., 2026).

70 Recent advances in natural language processing and machine learning have enabled the development of long-term multi-sectoral datasets (Dayrell et al., 2022) for systematic extraction of drought impact information from a large text corpus. However, ~~substantial uncertainty remains regarding how drought–impact relationships vary across accumulation timescales~~there are large differences in the variation of drought impact relationships across accumulation timescales, the role of atmospheric evaporative demand in shaping those relationships, and ~~how impact-reporting sensitivity changes with increasing drought severity~~the change in impact sensitivity with increasing drought severity. These gaps are particularly relevant for Spain, a region with amplified hydro-climatic gradients, high interannual variability and high sensitivity to both precipitation deficits and drought intensification driven by temperature (Dominguez-Castro et al., 2019; Vicente-Serrano et al., 2019; Noguera et al., 20222021). Despite extensive use of drought indices, the spatial organization of drought impacts and the dominant timescales controlling hydrological and agricultural responses remain poorly constrained. Addressing these

75 challenges is critical for improving impact-based drought monitoring and early warning systems, which increasingly require information on when, where, and how impacts occur. Spain provides an ideal case study due to its climatic heterogeneity and recurrent exposure to drought.

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The aim of this study is to quantify the multiscale relationship between drought severity (SPI and SPEI) and media-reported hydrological and agricultural impacts across Spain (1976–2023). Specifically, we (1) assess the temporal coherence and timescale dependence of drought–impact relationships, (2) evaluate the role of atmospheric evaporative demand (SPEI vs SPI) and the nonlinear sensitivity of impacts ~~reporting to drought severity~~to drought severity, and (3) identify the dominant timescales and spatial patterns ~~associated with impact-reporting responses across sectors~~controlling impact responses across sectors. The results provide new insights into ~~the temporal, nonlinear, and spatial dynamics of drought-impact reporting~~the temporal, nonlinear, and spatial dynamics of drought impacts and contribute to advancing impact-based drought monitoring

85

90 under changing climatic conditions.

2 Data and methods

2.1 Drought impact data

Drought impact information was extracted from El País, the general-interest newspaper with the largest circulation in Spain. All articles published from its foundation on 11 May 1976 to 31 December 2023 were analysed. Articles were initially pre-

95 filtered for drought relevance using a supervised classifier based on a fine-tuned Longformer model within the SeqIA framework (López-Otal et al., 2025). This binary classifier identified 15,102 drought-relevant news articles over the study period.

Drought impact extraction from these articles was conducted using a schema-guided Generative Information Extraction (GenIE) approach implemented within the CienaLLM framework (https://github.com/lcsc/ciena_llm/). This method enables
100 the extraction of structured information from unstructured text by prompting a large language model (LLM) to generate outputs (e.g., in JSON format) based on a predefined schema specifying the target information.

The first schema incorporated the four most common drought-impact categories in Spain: crops (e.g., reduction in yield of crops, crop failure, planting delays), livestock (e.g., reduced productivity, feed shortage, increased mortality), hydrological resources (e.g., reduced river discharge, reservoir depletion, decline in groundwater level) and energy (e.g., reductions in
105 hydropower generation). We did the extraction locally using the gemma2:9b model, aiming to balance computational cost and extraction performance. The extraction resulted in 6,852 articles reporting crop-related impacts and 1,268 articles reporting livestock-related impacts. These were eventually combined into one category of agricultural impacts comprising 7,009 articles. Hydrological resources (6,408 articles) and energy (1,106 articles) were merged into a broader category of hydrological impacts. These aggregations were introduced to reduce data sparsity and support consistent province-by-timescale comparisons
110 at a broad sectoral level. They represent operational groupings rather than an assumption that all constituent impacts share identical mechanisms or response timescales. The resulting sector-level patterns should therefore be interpreted as broad reporting signals.

Temporal attribution was based on the publication date of each article. Specifically, an article contributed to the calendar month in which it was published; temporal expressions within the article text were not used to back-date an impact to an
115 inferred occurrence month. The monthly series (1976–2023) therefore represent the timing of media reporting, which may differ from the timing of impact onset because of delays between impact occurrence, recognition, and publication.

Repeated coverage was intentionally retained at the article level. No event-level de-duplication was applied across consecutive articles or months, so multiple reports of an ongoing condition (for example, persistently low reservoir levels) contribute
120 multiple observations. Consequently, the resulting monthly series quantify reporting intensity or societal salience rather than the number of unique drought-impact events.

Province extraction was performed separately from impact extraction to identify the Spanish provinces referenced in drought-related articles and affected by drought~~Province extraction was performed to identify the Spanish provinces referenced in drought-related articles and affected by drought and was done separately from extracting impacts.~~ Because impact labels and
125 province labels were extracted independently, the workflow did not establish sentence-level one-to-one links between a

particular impact and a particular province when an article contained multiple impacts and multiple locations. Such articles can therefore contribute to more than one province–sector combination, and province-level results should be interpreted as spatial patterns of drought-related reporting rather than perfectly resolved geocoded impact events. The province names were standardized by mapping the different model outputs to the official names, reconciling spelling differences between co-official languages, and manually assigning frequently mentioned municipalities or geographical features to their corresponding provinces. The final dataset is a monthly count of articles referring to each impact category (agriculture and hydrology) in each province. Monthly impact-reporting frequencies were quantified as the number of drought-related articles published for each province and impact category. Article publication month was used for temporal attribution, and repeated reporting of persistent impacts was retained; consequently, the response variable represents reporting intensity and societal visibility rather than a count of unique impact events or a direct measure of physical damage magnitude. The extraction was multi-label rather than mutually exclusive. Therefore, a single article could contribute to more than one impact category when it reported multiple drought consequences, such as crop losses together with reservoir depletion or water-use restrictions. The agricultural and hydrological series should therefore be interpreted as sector-specific monthly reporting frequencies, not as disjoint sets of articles. To assess the possible influence of co-reporting, we quantified the article-level overlap between the two broad sectors and considered this overlap when interpreting the relationship between agricultural- and hydrological-impact reporting series.

2.2 Drought quantification

To compute drought indices, this study used a high-resolution gridded drought-index-dataset developed by Vicente-Serrano et al. (2017a) with SPI and SPEI information for the whole of Spain at a resolution of 1.1 km. SPI and SPEI, which are based on precipitation and on the difference between precipitation and atmospheric evaporative demand, respectively, have been widely used in agricultural, ecological and socioeconomic studies (e.g., Abatzoglou et al., 2014; Bachmair et al., 2015; Barker et al., 2016; Kumar-El Kenawy et al., 2016, 2026; Peña-Gallardo et al., 20182019). The monthly SPI and SPEI series were computed for the 48 provinces, including 47 mainland provinces and the Balearic Islands, of mainland Spain and for the Balearic Islands for the spatially consistent analysis, along but also an overall with overall national SPI and SPEI series. The SPI and SPEI were computed for accumulation periods from 1 to 48 months. While SPI is useful for its simplicity and availability of data (McKee and Doesken et al., 19932016), it does not take into account atmospheric demand and may underestimate the intensity of drought during warm periods, which is considered by the SPEI (Tomas-Burguera et al., 2020). The joint use of SPI and SPEI therefore offers a complete characterization of the drought severity and variability over Spain for the analysed period (1976–2023), useful for both climatological assessments and impact attribution.

2.3 Relating drought to impact data

To allow a robust comparison of the temporal dynamics of drought indices (SPI and SPEI) and the frequency-count of media reports on reported impacts-of-drought impacts, we standardized the time series of the monthly relative-frequency-count of drought-related news across the provinces of Spain for the period 1976–2023. The dataset includes information about-on

hydrological and agricultural drought impacts reported in newspapers. ~~Because the raw data have a substantial number of zeros in months with no identified impact reporting~~Because the raw data are proportional frequencies, which are often missing in ~~months with no reports~~, a statistical transformation was needed to normalize the distribution of the data and to make it comparable to standardized drought indices. We applied the method proposed by Vicente-Serrano et al. (2025), ~~which~~. ~~We concentrate on the application of~~applies a non-parametric empirical standardization technique for non-climatic impact indicators (e.g., media-based records) that may not satisfy normality or distribution assumptions. Specifically, it ~~it~~ deals with important issues of discrete and zero-inflated time series and provides a flexible yet rigorous framework to convert monthly impact-reporting frequencies counts into standardized values. This approach preserves the ordinal structure of the original series while accommodating repeated values and months without drought reports. This yields a standardized impact-reporting index directly comparable to SPI and SPEI, with negative values indicating below-average impact-reporting frequency counts (i.e., little or no reported drought activity) and positive values indicating above-average impact-reporting frequency counts, with increasing magnitude indicating increasing reporting frequency counts or visibility of drought conditions. This transformation placed all provincial time series on a common scale (mean = 0, standard deviation = 1), facilitating comparison among provinces and with SPI/SPEI. However, standardization does not remove possible long-term changes in newspaper production, editorial priorities, digital publication practices, or drought newsworthiness. These sources of non-climatic temporal variation are therefore treated as limitations of the media-derived record rather than assumed to have been eliminated by standardization. The analysis was contemporaneous: impact reporting in month ~~(t)~~ was compared with SPI or SPEI whose accumulation period ended in the same month ~~(t)~~. Thus, the analysis evaluates climatic accumulation windows associated with reporting rather than explicit temporal lags between drought conditions and subsequent impacts or reporting.

2.4 Spatial regionalization of drought-impact reporting data

To reduce dimensionality and identify dominant spatioal-temporal patterns in drought-related impact reporting, we applied Principal Component Analysis (PCA) to the standardized monthly drought-impact reporting time series. PCA was conducted separately for hydrological- and agricultural-impact reportings. In each sector-specific analysis, the input matrix consisted of monthly observations from 1976–2023 as rows ($n = 572$) and Spanish provinces as columns ($n = 48$). Therefore, the covariance matrix represented covariation among provincial impact-reporting time series within a given sector. SPI and SPEI were analysed subsequently in the drought-impact-reporting correlation and sensitivity analyses. The PCA was used only to identify the dominant spatial modes of impact-reporting reported impact-variability and the provinces contributing most strongly to those modes. ~~–~~Spatial patterns were used to visualize the loadings to recognize regions with similar behaviorbehaviour of drought-impact reportingdrought impacts. We examined the temporal evolution (scores) of each principal component to illustrate the dominant temporal mode of variability among provinces.

2.5 Drought–~~impact-reporting~~ sensitivity coefficient

To quantify how media reporting varies with drought magnitude, we calculated a drought–reporting sensitivity coefficient (β) from a log–log relationship between drought magnitude and monthly media-reporting counts. This coefficient provides a scale-independent measure of the elasticity of reporting intensity with respect to drought magnitude across sectors, provinces, drought-severity classes, and accumulation timescales.

For the sensitivity analysis, only months with dry index values (SPI or $\text{SPEI} < 0$) were retained, and the drought index was sign-reversed so that larger positive values represent greater drought magnitude ($D_t = -\text{SPI}_t$ or $-\text{SPEI}_t$). This means that the magnitude of drought was represented by the absolute value of the drought index (López-Moreno et al., 2013; Barker et al., 2016; Conradt et al., 2023). Unlike the standardized impact series used for the temporal comparison, correlation analysis, and PCA, the response variable in the log–log regression was the unstandardized monthly article count for the corresponding province and sector. Thus, I_t denotes the number of media reports in month t . Because $\ln(0)$ is undefined, months with $I_t=0$ were excluded from this specific regression step; months with positive article counts were transformed as $\ln(I_t)$, and drought magnitude was transformed as $\ln(D_t)$.

To complement this conditional log–log analysis and explicitly account for zero-report months, we also modelled the occurrence of drought-impact reporting using all dry months ($\text{SPI} < 0$ or $\text{SPEI} < 0$), including months with zero reports (i.e., no identified impact article). For each province, drought index, and accumulation timescale, reporting occurrence was represented as a binary response ($I_t > 0 = 1$; $I_t = 0 = 0$) and related to positive drought magnitude ($D_t = -\text{SPI}_t$ or $-\text{SPEI}_t$) using logistic regression. The coefficient γ , its 95% confidence interval, and the corresponding odds ratio $\exp(\gamma)$ were used to quantify the change in the probability of any reporting with increasing drought magnitude. This occurrence model complements, rather than replaces, the log–log elasticity coefficient β , which quantifies reporting intensity conditional on positive reporting.

To examine whether the reporting–drought association changes across drought severity, monthly observations were stratified by the sign-reversed drought index into ~~near~~-normal dry conditions (~~$0 < D_t \leq 0.5$~~), mild ($0.5 < D_t \leq -0.84$), moderate ($0.84 < D_t \leq -1.28$), severe ($1.28 < D_t \leq -1.65$), and extreme ($D_t > -1.65$) drought, together with a category containing all dry conditions ($D_t > 0$) (Agnew, 2000). Regressions were fitted separately within each class. This stratification is intended to reveal broad changes in reporting sensitivity across severity ranges, but the width of the predictor range and the number of observations differ among classes. In particular, severe and extreme classes ~~is-are~~ likely to include relatively few observations, so isolated class-specific slopes may be unstable. As such, the number of observations supporting each regression, as well as the resulting sample-size distributions, was quantified across provinces, accumulation timescales, drought indicators, sectors, and severity classes.

To evaluate the statistical precision of the severity-specific sensitivity estimates, we complemented the sample-size assessment with direct uncertainty diagnostics for β . For each province–index–timescale regression, we calculated the standard error and 95% confidence interval of β . Across provinces, uncertainty was summarized using the interprovincial distribution of β , the percentage of regressions whose 95% confidence intervals excluded zero, and the median 95% confidence-interval width. These diagnostics distinguish the amount of sample support from the precision with which the regression slope is estimated.

Overall, for each province, the ordinary least squares (OLS) estimator for β was computed, as follows:

$$\ln(I_t) = \alpha + \beta \ln(D_t) + \varepsilon_t$$

where I_t represents the monthly media-reporting count in month t , and D_t is the sign-reversed drought index at timescale t . Under the log–log formulation, the coefficient β quantifies the relative response of reporting frequency to changes in drought magnitude, indicating the proportional change in reporting intensity associated with a proportional change in drought magnitude (e.g., a 1% increase in drought magnitude is associated with an approximate $\beta\%$ change in reporting intensity). This statistic provides a dimensionless measure of drought-reporting sensitivity. As a result, comparisons can be made across sectors (hydrological vs. agricultural), provinces, and timescales (1–48 months). This enables a multiscale assessment of drought–impact–reporting relationships. Moreover, this framework enables the identification of the timescales, as well as provinces, at which drought conditions show the strongest association with reporting. For each sector, sensitivity coefficients were first estimated at the provincial level and then aggregated to the national scale by computing the arithmetic mean across provinces.

Importantly, prior to inferential interpretation of the correlation and regression analyses, temporal autocorrelation was evaluated for the drought-index and impact-reporting series. Serial dependence was assessed using the autocorrelation function (ACF), where autocorrelation of the regression residuals was additionally examined to verify that temporal dependence did not materially invalidate the statistical inference from the fitted models. To evaluate the role of atmospheric evaporative demand on drought severity and subsequent impact reporting, sensitivity coefficients were calculated independently using SPI and SPEI. Differences between SPI- and SPEI-based coefficients (i.e., SPI – SPEI) were used to assess the contribution of temperature-driven processes to drought-reporting responses. Negative differences between sensitivity coefficients (i.e., SPEI > SPI) indicate a stronger influence of atmospheric evaporative demand, whereas positive values (i.e., SPI > SPEI) suggest precipitation-dominated responses.

Regional patterns of impact-reporting response were determined by analysing the spatial distribution of drought-reporting impact-sensitivity at the province level. In addition, the dominant accumulation timescale was defined as the timescale for which the sensitivity coefficient was the maximum in the range 1–48 months for each province and drought severity class. All drought indices were aligned with media reporting in the same month; therefore, the identified accumulation timescale represents the climatic integration window most strongly associated with reporting at time t , rather than drought-event

duration or an explicit impact-response ~~lag-lag~~. In cases where multiple timescales exhibited similar values, the smallest timescale among the maxima was selected to ensure consistency.

3. Results

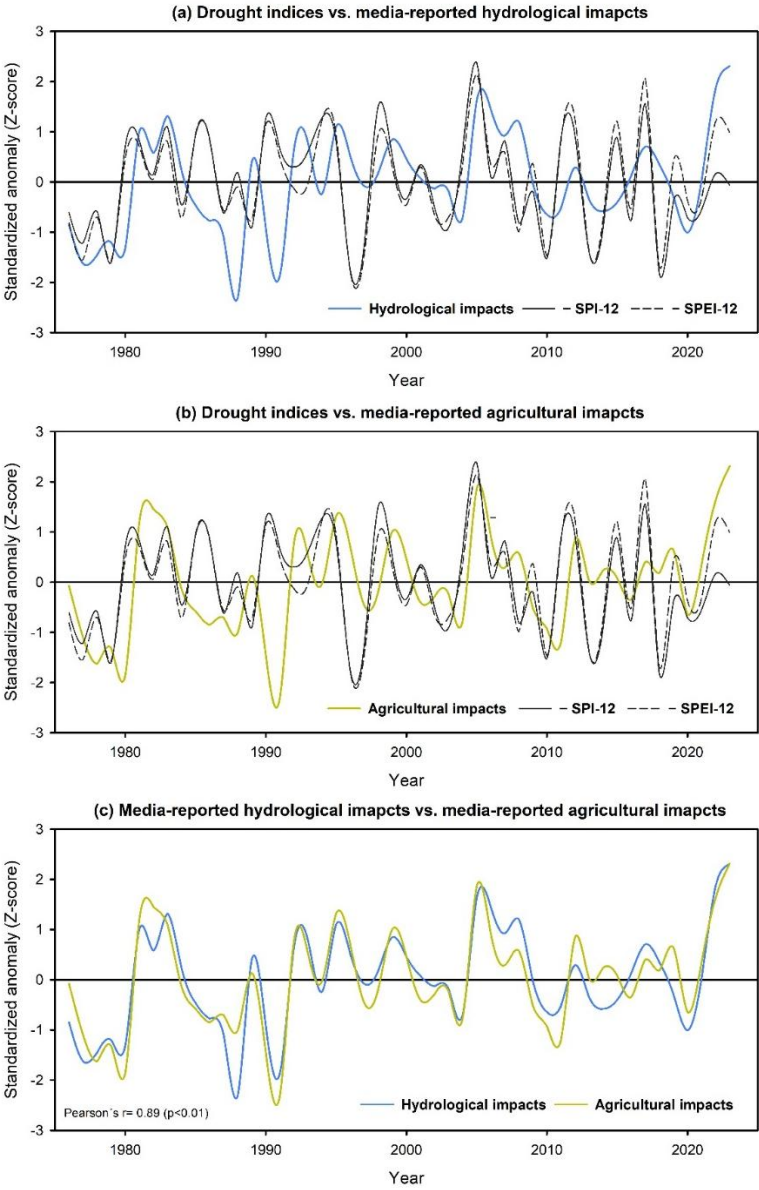
Throughout this section, the terms hydrological and agricultural impacts are used as shorthand for the corresponding media-derived impact-reporting series. Accordingly, reported sensitivities, spatial patterns, and dominant timescales refer to variation in the frequency and intensity of reporting on these impacts, rather than direct changes in physical impact magnitude or the number of unique impact events.

3.1. Drought impacts ~~reporting~~ tracks climate variability across timescales

Figure 1 depicts the temporal evolution of standardized drought indices (SPI-12 and SPEI-12) and media-reported drought impacts in Spain from 1976 to 2023. Hydrological reporting closely tracks ~~multi-year drought variability~~, with major ~~peaks in drought frequency impacts~~ ~~drought-impact reporting~~ occurring in the early 1980s, mid-1990s, early 2000s and early 2010s associated with significant positive anomalies. In contrast, wet periods are associated with negative anomalies (< -1), indicating a decrease ~~in or absence of impact reporting~~ ~~no-reported impacts~~. Although both sectors exhibit coherent variability, important sectoral differences emerge. Negative SPEI anomalies have become more frequent than SPI anomalies in recent decades (after 2000) and are more consistent with variability in reporting. However, some inconsistencies are observed, such as the ~~finding that the effects of mild-to-moderate drought conditions~~ (-0.5 to -1) are not always significant, indicating nonlinear and context-dependent responses. The strong relationship between agricultural- and ~~hydrological-hydrological-impact reporting series~~ ~~impacts~~ ($r = 0.89$, Fig. 1c) should be interpreted with caution because the impact categories are not mutually exclusive at the article level. Some articles report multiple drought consequences simultaneously and therefore contribute to both sectoral series. Accordingly, Fig. 1c reflects both genuine temporal co-occurrence of sectoral impacts during major drought episodes and partial co-reporting within the same media articles. We therefore interpret this relationship as coherent cross-sector drought reporting and impact co-occurrence, rather than as fully independent evidence of causal propagation between sectors. τ

The relationship between drought indices and ~~impact reporting~~ ~~impacts~~ across accumulation timescales (1–48 months) over Spain is summarized in Fig. 2. Consistent negative correlations indicate that ~~drier conditions correspond to increased reported impacts~~ ~~drier conditions correspond to increased impact reporting~~. Correlations increase significantly with increasing timescale, reaching maximum values at 12–30 months for both sectors ($|r| \approx 0.6$ – 0.8), while short timescales (1–3 months) indicate weak and inconsistent relationships ($|r| < 0.3$). These relationships are further modulated by seasonal variability, with higher correlations during ~~summer and months of summer and~~ early autumn when water demand and agricultural sensitivity peak. Hydrological-~~impact reporting~~ ~~impacts have~~ ~~has~~ the highest correlations at longer time-scales (12–32 months), whereas agricultural-~~impact reporting~~ ~~impacts~~ responds mainly over a slightly broader range, especially in late spring and early summer. SPEI has slightly stronger correlations than SPI across most conditions and especially for agriculture. The results indicate that

290 impact-reporting reported impact frequency is more strongly associated with persistent moisture deficits than with short-term anomalies.



295 **Fig. 1: Temporal evolution of standardized drought indices and media-reported drought impacts in peninsular Spain over the period 1976–2023. (a) Hydrological-impact reportingHydrological impacts and (b) agricultural-impact reporting agricultural impacts are compared with SPI-12 and SPEI-12, while (c) illustrates the relationship between hydrological-impact reporting and agricultural-impact reportinghydrological and agricultural impacts. All series are expressed as standardized anomalies (Z-scores). In panels a and b, drought indices were sign-reversed (multiplied by -1) to facilitate direct comparisons with drought-impact reporting drought impacts (i.e., drought severity $\uparrow$, impact reportingimpacts $\uparrow$).**

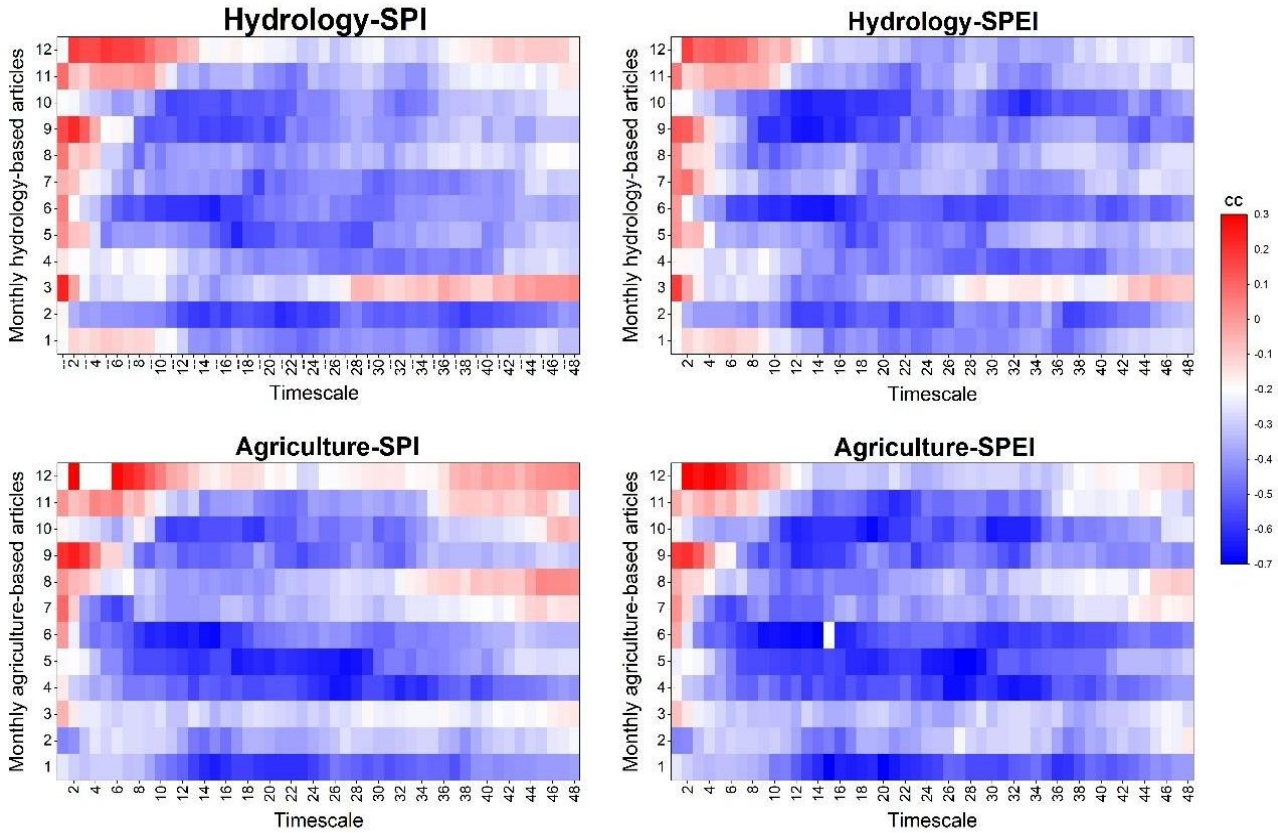


Fig. 2: Correlation matrix (Pearson coefficient; **CC**) between standardized monthly frequencies of newspaper reports on drought impacts ~~drought-related newspaper impacts~~ and drought indices (SPI and SPEI) computed at accumulation timescales from 1 to 48 months. Absolute values of **CC** exceeding 0.28 are statistically significant at $p < 0.05$.

3.2. Mechanisms and nonlinear sensitivity of drought-reporting drought impact relationships

Figure 3 illustrates the drought-reporting sensitivity coefficient of media-reporting frequency to drought severity across accumulation timescales. Reporting sensitivity varies markedly with timescale and drought intensity, and the strongest response coefficient is found at 12–24 months. The sensitivity coefficient values are consistently larger for SPEI compared to SPI, especially for hydrological impacts ~~especially for hydrological impact reporting~~, indicating that temperature-driven processes are associated with stronger reporting responses ~~amplify impact responses~~. While the sensitivity coefficient for hydrological impact reporting ~~hydrological impacts~~ indicates smoother, more gradual increases, agricultural reporting displays higher and more variable sensitivity, which reflects ~~higher and more variable sensitivity, which may reflect~~ more rapid changes ~~in agricultural impact reporting in their faster reaction response~~ to soil moisture deficits. At short timescales (1–3 months), the sensitivity coefficient is low, confirming the limited role of short-term anomalies. These findings highlight the nonlinear response of the frequency of media reports on impacts ~~media reported impact frequency~~ to drought severity and the dominant role of sustained moisture deficits.

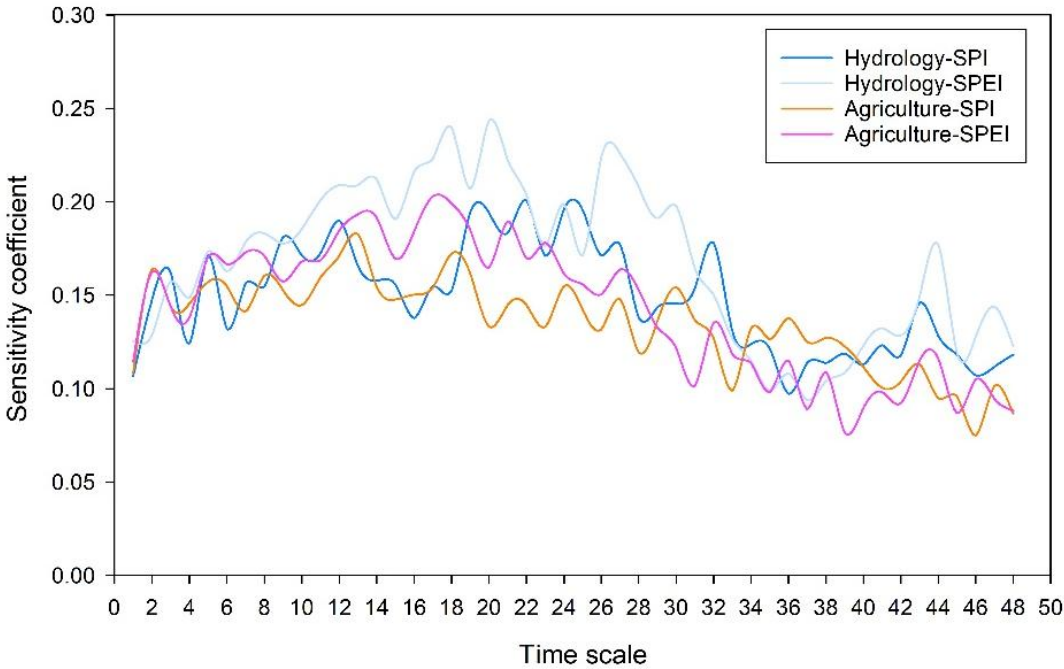


Fig. 3: ~~Drought-reporting~~ ~~Drought-impact~~ sensitivity coefficient response curves linking drought indices (SPI and SPEI) with ~~media-reporting frequencies for media-reported hydrological and agricultural drought impacts across accumulation timescales~~ hydrological- and agricultural-impact reporting across accumulation timescales from 1 to 48 months. Lines represent mean ~~drought-reporting drought-impact~~ sensitivity coefficient values averaged across Spanish provinces for each sector-index combination.

~~Because stratification by drought severity class reduces the number of observations available for individual regressions, we examined the sample size underlying the sensitivity estimates across sectors, drought indicators, and accumulation timescales (Supplementary Figs. S1-S3). As expected, for the hydrological (Supplementary Fig. S1) and agricultural impacts (Supplementary Fig. S2), the number of observations decreases from normal and moderate conditions toward severe and extreme drought classes. Nevertheless, the severe and extreme regressions retain observations across the full 1–48 month accumulation range for both SPI and SPEI and for both hydrological and agricultural impacts. Median sample sizes generally remain around or above 10 observations across most timescales, and frequently approach or exceed 15 observations over the intermediate accumulation periods where many of the strongest sensitivity responses occur (Supplementary Fig. S3). The largest reductions in sample size are more evident toward the longest accumulation periods, particularly for some extreme drought combinations. The interquartile ranges further indicate appreciable spatial variability in the number of observations available among provinces, confirming that some province timescale regressions are less well supported than others.~~

Importantly, however, the sample size curves vary smoothly with accumulation timescale rather than showing isolated collapses at the accumulation periods associated with the principal sensitivity maxima. This suggests that the broad timescale-dependent patterns are not simply artifacts of a single or very small number of sparsely sampled regressions.

Because stratification by drought-severity class reduces both sample size and predictor range, we evaluated the support and statistical precision of the sensitivity estimates using complementary diagnostics (Supplementary Figs. S1–S4). Sample sizes decline from normal and moderate toward severe and extreme drought, particularly at the longest accumulation periods, although severe and extreme regressions retain observations across the full 1–48-month range and median sample sizes generally remain around or above 10 observations across most timescales (Supplementary Figs. S1–S3). However, sample size alone does not ensure precise slope estimation. Direct uncertainty diagnostics show substantial interprovincial variability in β , a generally modest proportion of province-level regressions whose 95% confidence intervals exclude zero, and broader confidence intervals under severe and extreme drought conditions (Supplementary Fig. S4). These results indicate that the broad timescale-dependent response is more robust at the aggregated national scale than for individual province–timescale coefficients, which should therefore be interpreted cautiously.

To further evaluate the sensitivity-analysis assumptions, we complemented the conditional log–log analysis with a zero-inclusive occurrence model that retains months with no reported impacts (Supplementary Fig. S5). This analysis separates the probability that reporting occurs at all from the reporting intensity conditional on positive reporting, represented by β . The zero-inclusive results show that greater drought magnitude is generally associated with higher odds of reporting, while the strength of this occurrence response varies across accumulation timescales. The strongest and most spatially coherent occurrence responses are generally concentrated at short-to-intermediate timescales, particularly for SPEI, with a substantial proportion of province-level logistic coefficients showing 95% confidence intervals that exclude zero over these ranges.

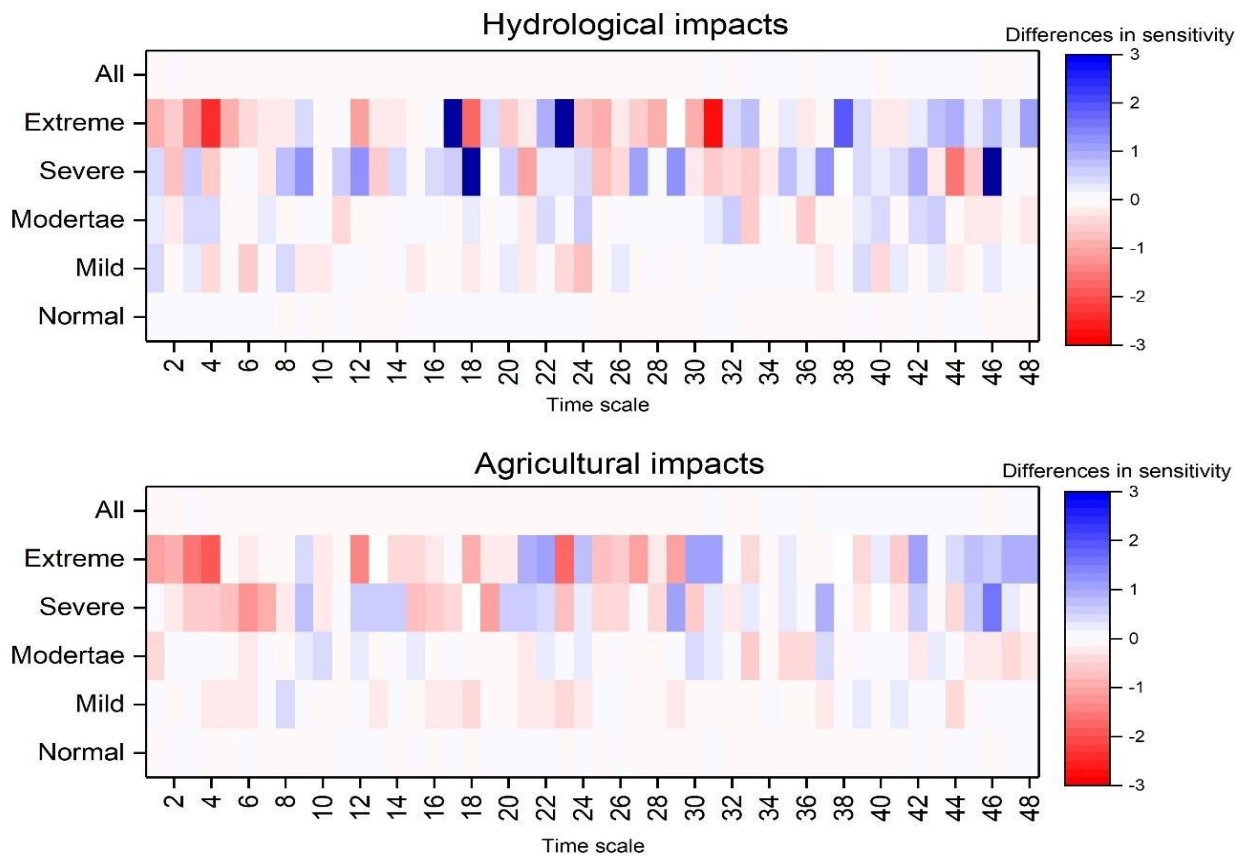


Fig. 4: Differences in national-scale mean ~~drought-reporting drought-impact~~ responses between SPI and SPEI across accumulation timescales (1–48 months). Differences between SPI and SPEI responses indicate the contribution of temperature-driven evaporative demand across drought severity classes.

Results suggest that ~~reporting impact~~ sensitivity is dependent on accumulation timescale (Fig. 4), with the highest sensitivity at 12–36 months. Differences between SPEI- and SPI-based responses are negligible at short timescales (1–3 months), but increase with increasing timescales, with consistently stronger responses from SPEI, particularly for moderate to severe droughts. Results reveal some sectoral differences. Specifically, for ~~hydrological-impact reporting hydrological impacts~~, negative differences emerge primarily at medium- to long timescales (12–36 months). In contrast, ~~agricultural-impact reporting agricultural impacts suggest shows~~ earlier divergence, with amplified negative differences already at shorter- to intermediate timescales (6–24 months), indicating higher sensitivity to atmospheric demand.

Figure 5 reveals clear spatial heterogeneity in drought-~~reporting impact sensitivities~~ sensitivity. SPEI-dominated responses prevail over large areas, especially at longer timescales (12–36 months), ~~indicating a widespread influence of atmospheric evaporative demand on drought impacts. This effect is especially evident for agricultural systems, whereas hydrological~~

responses exhibit indicating that atmospheric evaporative demand is associated with stronger drought-reporting responses. This association is especially evident for agricultural-impact reporting, whereas hydrological-impact reporting exhibits greater spatial coherence at longer accumulation periods. At shorter timescales (3–6 months), differences are weaker and more fragmented, reflecting the dominant role of precipitation variability. Collectively, these results indicate a transition from precipitation-driven to temperature-influenced drought impacts consistent with a greater role of precipitation variability. Collectively, these results indicate a transition from precipitation-dominated to increasingly temperature-influenced drought-reporting responses at longer accumulation timescales.

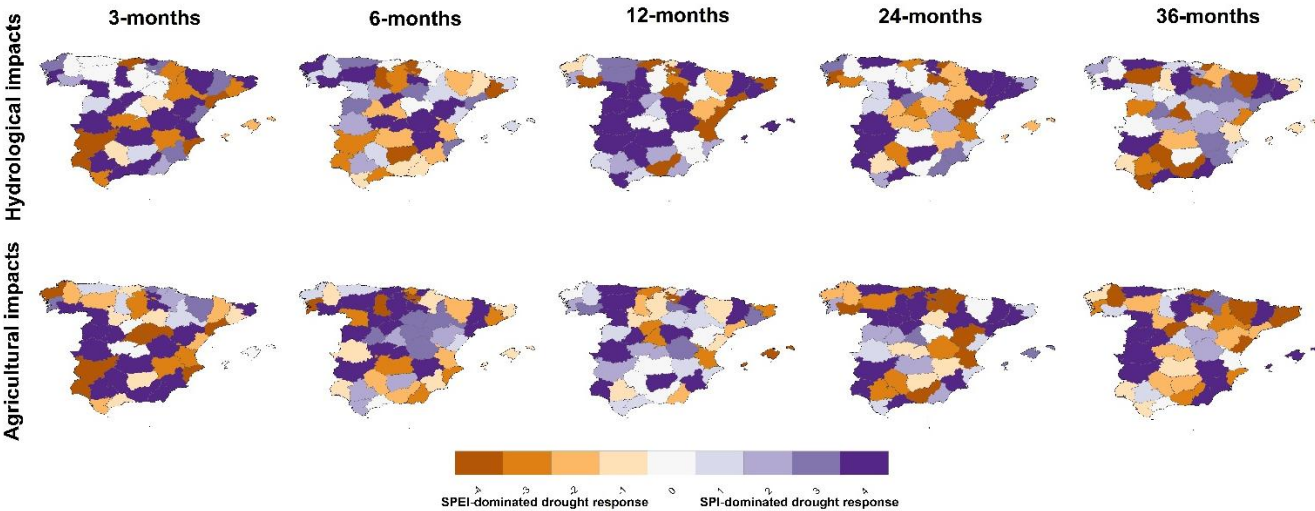


Fig. 5: Spatial differences in drought-reporting drought-impact sensitivity between SPI and SPEI across peninsular Spain at multiple accumulation timescales. The relative dominance of SPEI-driven versus SPI-driven drought reporting responses for hydrological-impact reporting hydrological impacts (top row) and agricultural-impact reporting agricultural impacts (bottom row) is presented at 3-, 6-, 12-, 24-, and 36-month timescales.

The leading spatial modes of drought-impact reporting drought impacts obtained from principal component analysis are illustrated in Fig. 6. The first principal component (PC1) embodies a prevailing national mode in the two sectors with positive loadings indicating coherent large-scale variability of drought-impact reporting drought impacts. The most evident hydrological impacts are in north-western and central Spain, consistent with basin-scale water storage responses, while agricultural impacts are more important in western and south-western regions, reflecting sensitivity to soil moisture deficits. PC2 demonstrates a clear east-west contrast with the Mediterranean and southern regions having stronger loadings and the interior and north having weaker or opposite patterns, a contrast that is more sharply defined for hydrological impacts. PC3 The strongest hydrological-impact reporting loadings occur in northwestern and central Spain, while agricultural-impact reporting is more prominent in western and southwestern regions. These spatial patterns are consistent with differences in basin-scale water-storage responses and sensitivity to soil-moisture deficits. PC2 demonstrates a clear east-west contrast, with the Mediterranean and southern

regions having stronger loadings and the interior and north having weaker or opposite patterns; this contrast is more sharply defined for hydrological-impact reporting. PC3–PC5 are higher-order components that represent more local variation. Hydrological impacts seem to show structured regional differences, mainly related to catchment characteristics, whereas agricultural impacts are more patchy and spatially heterogeneous. Overall, hydrological impacts are more spatially coherent while order components that represent more localized variation. Hydrological-impact reporting shows more structured regional differences, whereas agricultural-impact reporting is more spatially heterogeneous. Overall, hydrological-impact reporting is more spatially coherent, whereas agricultural-impact reporting agricultural impacts is are more locally modulated.

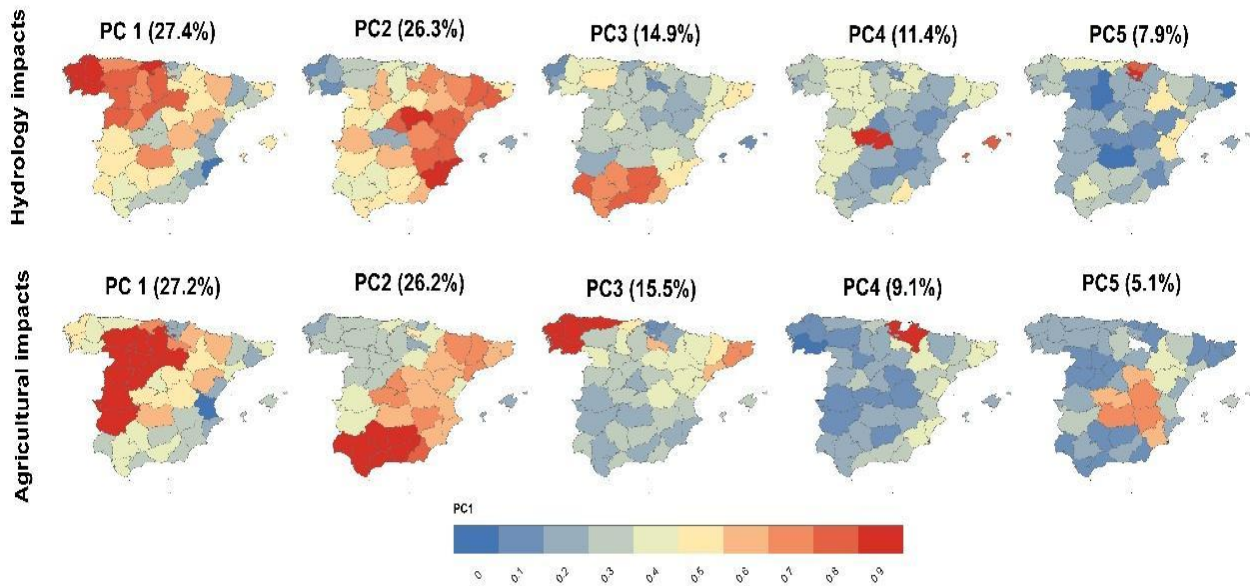
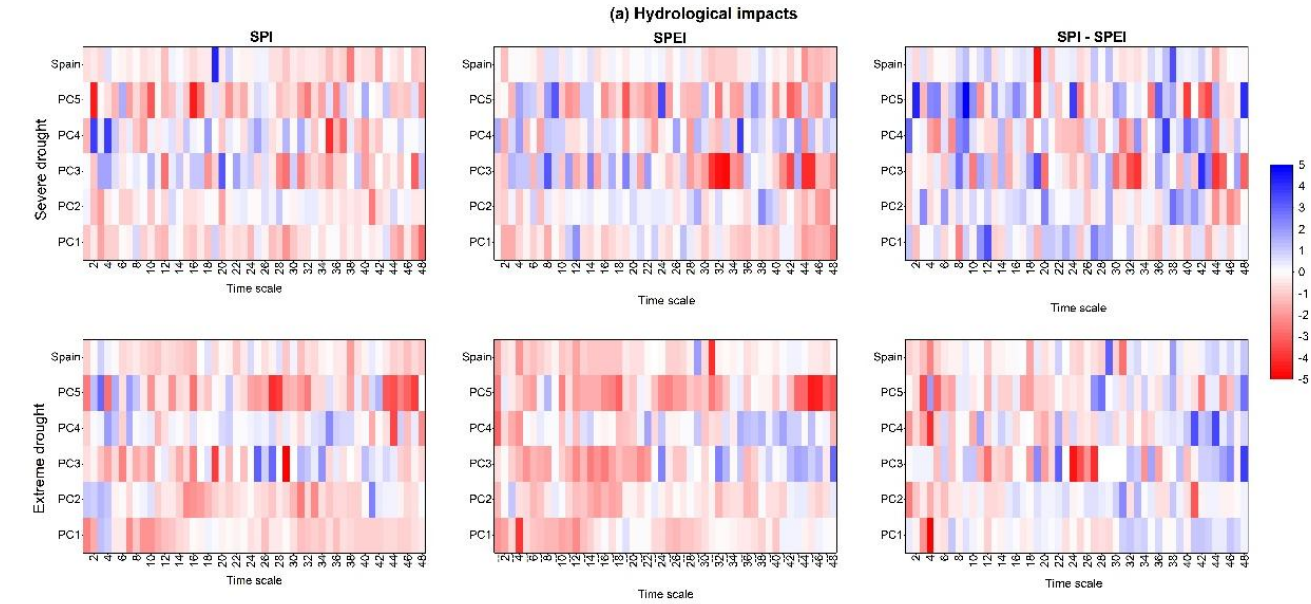


Fig. 6: Spatial loadings of the principal components (PCs) derived from standardized monthly hydrological (upper row) and agricultural (lower row) ~~drought-impact reporting drought-impact~~ time series across Spanish provinces (1976–2023). Numbers in parentheses indicate the percentage of total variance explained by each principal component.

The ~~Dd~~ drought-reporting sensitivity coefficient was assessed under different drought severity classes sensitivity coefficient was assessed under different drought-severity classes for the different PCs in both hydrological- and agricultural-impact reporting for both hydrological and agricultural impacts (Fig. 7). As illustrated, for hydrological-impact reporting-impacts, sensitivity increases with both drought severity and accumulation period, peaking during severe and extreme droughts at 12–36 months. SPEI consistently exhibits larger drought-reporting-impact sensitivity coefficients than SPI, especially under extreme conditions, suggesting an increased sensitivity to atmospheric evaporative demand. At the regional scale, PC1, the dominant large-scale signal over northwestern and central Spain, exhibits a growing sensitivity with timescale and drought severity reaching its maximum exhibits increasing sensitivity with timescale and drought severity, reaching its maximum at 12–36

415 months, ~~in agreement with the integrative~~ which is consistent with the integrative behaviour of large river basins and storage
systems. ~~The PC2, which is related to the Mediterranean and southern regions, is more sensitive to severe and extreme drought,~~
~~especially at the longer timescales, implying a higher vulnerability to prolonged moisture deficits in these regions.~~ PC2, which
is related to the Mediterranean and southern regions, is more sensitive to severe and extreme drought, especially at longer
timescales, indicating stronger reporting sensitivity to prolonged moisture deficits in these regions. In contrast, higher-order
420 components (PC3-5) that represent more localized variability have more irregular and spatially restricted responses with the
~~sensitivity peaks differing by timescale-order components (PC3-PC5), which represent more localized variability, have more~~
~~irregular and spatially restricted responses, with sensitivity peaks differing by timescale.~~



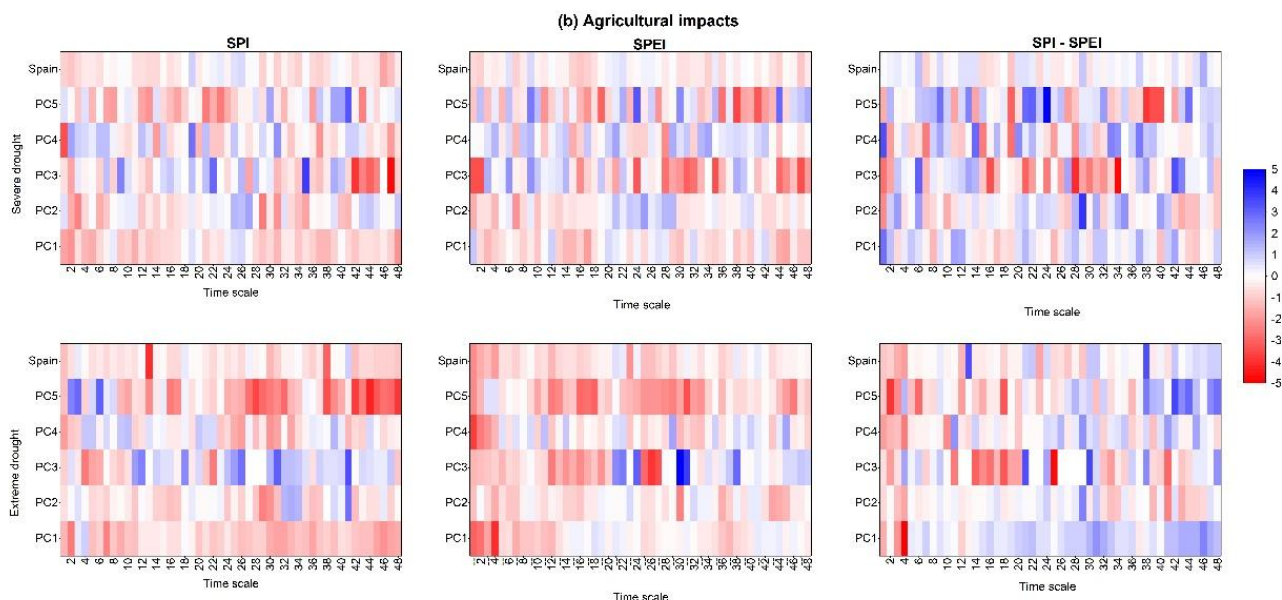


Fig. 7: Drought-reporting impact sensitivity coefficient between drought indices (SPI and SPEI) and (a) hydrological- hydrological and (b) agricultural-impact reporting agricultural drought impacts across accumulation timescales (1–48 months) under severe and extreme different drought severity classes drought classes. Panels show coefficient patterns for severe and extreme drought conditions derived for the main principal components (PC1–PC5), as well as peninsular Spain. Warmer colours indicate stronger responsiveness of impact reporting reported impacts to drought severity. Differences between SPI and SPEI highlight the influence of temperature-driven evaporative demand under extreme drought conditions.

Agricultural reporting shows higher sensitivity, with the peak drought-reporting drought impact sensitivity coefficient occurring at shorter-to-medium timescales short-to-medium timescales (6–24 months) (Fig. 7). In addition to the hydrological As with hydrological-impact reporting responses, the drought-reporting impact sensitivity coefficient sharply increases from moderate to extreme drought conditions, considering the higher vulnerability of agricultural systems consistent with stronger reporting responses under severe conditions. SPEI reveals stronger responses than SPI across all severity classes, again highlighting the importance of temperature-driven stress. For agriculture agricultural-impact reporting impacts (Fig. 7), there are also marked sub-regional subregional differences. The first principal component (PC1, west and southwest) exhibits high and fast increases in sensitivity at short to intermediate time rapid increases in sensitivity at short-to-intermediate time scales (6 to 24 months). PC2 and PC3, linked to eastern and northern areas, are characterized by a greater variability scales (6–24 months). PC2 and PC3, linked to eastern and northern areas, are characterized by greater variability and earlier sensitivity peaks. Higher-order components (PC4–PC5) have very local and variable responses with steep increases under extreme drought conditions indicating robust local modulation by soil properties, irrigation practices and crop type show localized and variable responses, with steep increases under extreme drought conditions, suggesting possible local modulation by soil properties, irrigation practices, and crop type. These results demonstrate that SPEI yields higher sensitivity than SPI for all PCs and the differences tend to increase for severe and extreme drought. However, this effect tends to be more spatially

structured ~~in hydrological systems and more fragmented in agricultural ones~~ in hydrological-impact reporting and more fragmented in agricultural-impact reporting.

3.3. Dominant drought accumulation timescales ~~associated with impacts~~ associated with impact reporting

450 Figures 8 synthesizes the dominant accumulation timescales associated with maximum ~~reporting impact~~ sensitivity. Specifically, the spatial distribution of the dominant drought accumulation timescale associated with maximum ~~drought-~~ ~~reporting drought impact~~ sensitivity across Spanish provinces for ~~hydrological- and agricultural-impact reporting~~ hydrological ~~and agricultural impacts~~ is presented for different ~~drought severity classes~~ drought-severity classes (normal, mild, moderate, severe, extreme, and all dry conditions). A clear sectoral contrast emerges. ~~Hydrological~~ Hydrological-impact reporting impacts ~~are is~~ predominantly ~~associated with controlled by~~ medium- to long-term timescales (12–48 months), ~~but with a widespread dominance~~ with widespread dominance of 12–24 months under moderate conditions and a shift toward longer timescales (24–48 months) under severe and extreme drought. In contrast, ~~agricultural agricultural-impact reporting impacts are is~~ mainly associated with ~~shorter short to intermediate timescales to intermediate timescales~~ (3–12 months), particularly under normal to moderate drought, consistent with stronger associations at shorter accumulation timescales and ~~sensitivity to soil moisture~~ agricultural sensitivity to soil-moisture deficits (Fig. 8). ~~However, in extreme conditions, agricultural systems suggest a partial shift towards longer timescales, but still shorter than hydrological responses. Under extreme conditions, however, agricultural-~~ impact reporting shows a partial shift toward longer timescales, although these remain shorter than those associated with ~~hydrological-impact reporting~~. Hydrological ~~reporting~~ patterns are spatially more homogeneous, ~~while agricultural whereas~~ ~~agricultural reporting impacts demonstrates shows~~ greater regional variability, ~~pointing to a stronger local modulation pointing~~ to stronger local modulation. SPEI exhibits, in general, ~~longer dominant timescales than SPI in both sectors, especially for severe and extreme droughts~~ SPEI generally exhibits longer dominant timescales than SPI in both sectors, especially under severe and extreme drought, indicating stronger SPEI-based associations at longer accumulation timescales. ~~This effect is more pronounced for agricultural impacts, especially for severe and extreme drought, This effect is more pronounced for agricultural-~~ impact reporting, particularly under severe and extreme drought, where SPEI is associated with a shift toward longer dominant

460

465

470 accumulation timescales.

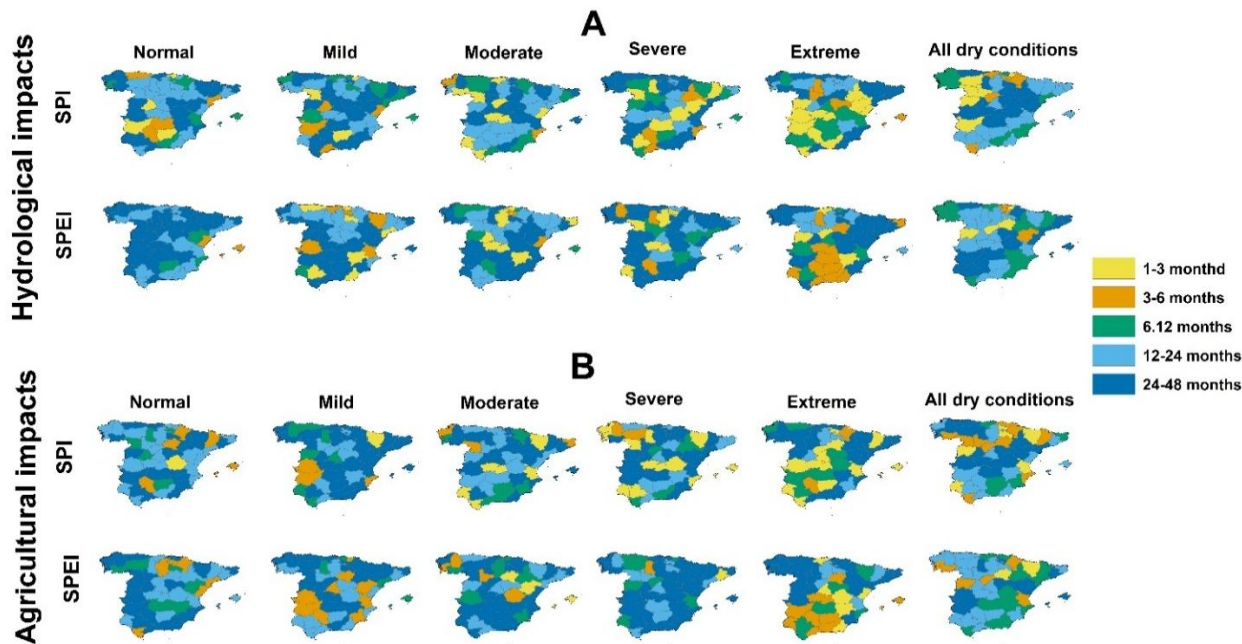


Fig. 8: Spatial distribution of the dominant drought accumulation timescale corresponding to the maximum impact sensitivity coefficient between drought indices (SPI and SPEI) and (a) hydrological and (b) agricultural drought impacts across Spanish provinces. Each panel corresponds to a drought severity class (normal, mild, moderate, severe, extreme, and all dry conditions). Colors represent the timescale at which drought conditions show the strongest association with impact reporting, ranging from meteorological drought (1–3 months) to long-term water storage deficits (24–48 months).

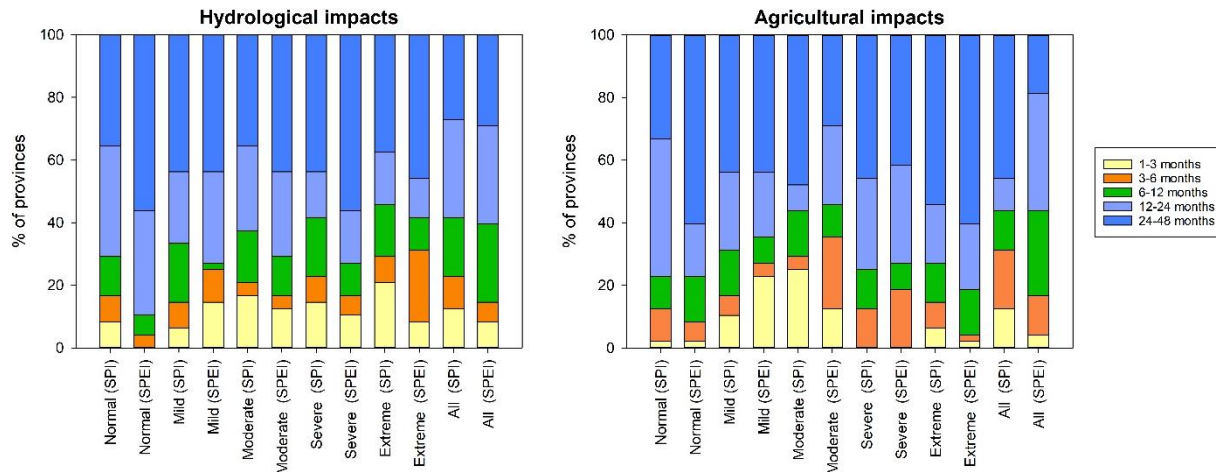


Fig. 9: Percentage of Spanish provinces associated with the dominant drought accumulation timescale derived from the maximum impact sensitivity between drought indices (SPI and SPEI) and reported impacts. Stacked bars show the distribution of provinces across five drought timescale categories, meteorological (1–3 months), agricultural (3–6 months), seasonal (6–12 months), hydrological (12–24 months), and long-term storage (24–48 months), for different drought severity classes (normal,

mild, moderate, severe, extreme, and all dry conditions). Results are shown separately for hydrological-~~impact reporting impacts~~ (left) and agricultural-~~impact reporting impacts~~ (right).

These patterns are summarized in Fig. 9, which indicates the percentage of Spanish provinces associated with the dominant drought accumulation timescale obtained from the maximum ~~drought-reporting drought-impact~~ sensitivity coefficient between drought indices (SPI and SPEI) and ~~impact reporting reported impacts for hydrological and agricultural sectors, for the hydrological and agricultural sectors.~~ Figure 9 shows that most provinces fall within the 12–48-month range ~~for agricultural and hydrological impacts, for agricultural- and hydrological-impact reporting.~~ However, agricultural ~~reporting impacts shows indicate~~ a higher concentrations ~~withinfor the 1–6 month1–6-month~~ range. ~~Short-lived anomalies, on the other hand, have a limited impact, as the short timescales (1–6 months) contribute little to the hydrological impact.~~ Short-lived anomalies show limited associations with hydrological-impact reporting, as short timescales (1–6 months) contribute little to the hydrological ~~reporting responses.~~ With increasing drought severity, both sectors indicate a shift towards longer dominant timescales, but this shift is more pronounced for agricultural systems. However, SPEI shifts the distribution to larger dominant timescales than SPI in all sectors consistently, especially in severe and extreme drought categories. This difference remains limited under mild drought conditions but increases substantially during severe and extreme droughts, indicating stronger SPEI-based associations at longer accumulation timescales. The effect is particularly striking for agricultural-~~impact reporting impacts~~, where SPEI enhances the role of ~~seasonal to long-term timescalesseasonal-to-long-term timescales~~ (6–48 months) while diminishing the contribution of short-term scales (1–6 months).

~~In hydrological systems~~For hydrological-impact reporting, SPEI further supports the dominance of long storage timescales (24–48 months), reflecting cumulative water deficits. These results indicate that SPEI shifts the dominant associations toward longer accumulation timescales. Collectively, these results demonstrate ~~a~~ systematic shift from shorter to longer dominant accumulation timescales ~~with increased severity and highlight contrasting temporal sensitivities of hydrological and agricultural impacts, with increasing drought severity and highlight contrasting temporal sensitivities of hydrological- and agricultural-impact reporting.~~

4. Discussion

This study provides a comprehensive assessment of the links between drought severity and ~~impact reporting observed impacts~~ across Spain through the combination of multiscale drought indices (SPI and SPEI) based on a high-resolution gridded climate

520 dataset and standardized media-reported hydrological and agricultural impacts for the period 1976–2023. This study reveals the temporal, nonlinear, and spatial dynamics of ~~drought-impact reporting drought impacts~~ through a scale-independent measure of ~~impact-reporting~~ response, providing new insights into drought-reporting relationships across accumulation timescales and sectors.

525 The multiscale analysis reveals weaker associations with short-term anomalies, whereas medium- and long-term accumulation periods show stronger associations with both hydrological and agricultural reporting. This pattern aligns with growing evidence that the impacts of drought are a function of the accumulation and duration of moisture deficits (e.g., [Van Loon and Laaha, 2015](#); Peña-Gallardo et al., 2019; ~~Orth-Altm~~ et al., ~~2021~~2020). ~~For example,~~ Peña-Gallardo et al. (2019) observed stronger relationships between meteorological and hydrological droughts at larger timescales in U.S. basins. Aryal et al. (2024) also
530 demonstrated an increase in the duration of propagation of droughts with the accumulation period, which implies an increase in system memory. These findings are consistent with earlier studies that have identified drought duration and persistence as important impact drivers (Orth et al., 2020; Álvarez-Garretón et al., 2021; Tonetto et al., 2024). Our results in Spain also suggest that hydrological reporting is more strongly associated with longer accumulation periods, whereas agricultural reporting shows stronger associations with intermediate accumulation periods, consistent with their differing dependence on
535 accumulated water deficits, soil moisture, and seasonal variability. The analysis indicates that isolated precipitation deficits are less strongly associated with ~~impact reporting media-reported impacts~~ than persistent deficits. Importantly, although newspaper reports often describe drought consequences when they become observable or socially salient, the temporal delay between physical impact occurrence and publication was not quantified in this study. For many reported impacts, this delay may be relatively short compared with the multi-month climatic accumulation windows examined here; however, drought
540 impacts are often gradual and persistent, and some reports may describe conditions that began earlier. We therefore use publication month as the timing of reporting rather than as an exact estimate of impact onset. Accordingly, any temporal offset between climatic conditions and the media record may reflect a combination of physical drought-impact response time, recognition of the impact, and reporting delay.

545 The study demonstrates that SPEI is more strongly associated with media-reported impacts than SPI, consistent with an important role of atmospheric evaporative demand in the determination of drought-reporting relationships. This contrast becomes more evident during prolonged and severe droughts. This suggests that temperature-driven processes are magnifying the effect of precipitation deficits, especially in the case of prolonged drought. This finding is in agreement with other global (Woodhouse et al., 2016; Ogunrinde et al., 2026), Mediterranean (Dong et al., 2019; Avanzi et al., 2020) and Spanish research
550 (Noguera et al., 2021). Atmospheric evaporative demand is particularly important for agriculture because increased atmospheric demand intensifies plant water stress and induces a large moisture loss from soil (Rigden et al., 2020; Qing et al., 2023). Hydrological systems respond more gradually and cumulatively, leading to long-term reductions in groundwater recharge and water storage. The higher atmospheric demand can enhance agricultural and ecological droughts (Vicente-

Serrano et al., 2022). Terrestrial water storage is strongly controlled by imbalances between water supply and atmospheric demand (Chang et al., 2020). The consistently stronger performance of SPEI relative to SPI further suggests that future droughts will increasingly reflect the combined effects of precipitation deficits and enhanced atmospheric evaporative demand associated with regional warming. This finding is particularly important in the Mediterranean basin, identified as a climate-change hotspot where temperatures are projected to rise faster than the global average, accompanied by increased drought severity and reduced water availability (Lionello and Scarascia, 2018).

Our findings further demonstrate a marked nonlinear drought-reporting relationship, with reporting sensitivity increasing disproportionately under severe and extreme drought. Nonlinear responses have been well documented for many regions and systems (Lobell et al., 2014; Chen et al., 2022; Wu et al., 2025; Liu et al., 2025). These findings suggest a threshold-like increase in reporting sensitivity; however, this should not be interpreted as direct evidence that physical impact magnitude increases at the same rate. Hydrological systems respond more slowly but exhibit stronger cumulative sensitivity, possibly related to the buffering effects of groundwater storage and reservoir regulation. However, studies in Spain indicate that these buffering mechanisms are insufficient to offset severe drought impacts. Previous studies have indicated that streamflow and water storage are highly sensitive to climate variability in Spanish basins (Berghuijs et al., 2016) and that reservoir regulation continues to alter but not offset the frequency and severity of droughts (Vicente-Serrano et al., 2014, 2017b). Moreover, more reliance on water storage and transfers increases vulnerability (Lorenzo-Lacruz et al., 2012, 2013) and risks of drought are high despite better management (Vargas and Paneque, et al., 2019). On the other hand, agricultural systems exhibit stronger and more variable increases in exposure with a limited ability to buffer this shift. These findings highlight the nonlinear amplification of the frequency of media reports on impacts media-reported impact frequency and underscore the need to incorporate nonlinear dynamics into drought risk assessments.

The supplementary robustness analyses qualify the interpretation of the drought-reporting sensitivity coefficient (Supplementary Figs. S4–S5). Although coherent timescale-dependent patterns remain evident after aggregation across provinces, individual province-level β estimates are not uniformly well constrained, particularly under severe and extreme drought where severity stratification narrows the predictor range and confidence intervals broaden. The principal multiscale patterns should therefore be interpreted more strongly at the aggregate sectoral scale than as precise estimates for individual province-timescale combinations. The zero-inclusive analysis further shows that drought reporting comprises two related but distinct components: the initiation of reporting and the subsequent intensity of reporting once it occurs. Greater drought magnitude is generally associated with higher odds of any reporting, with the most spatially coherent occurrence responses concentrated at short-to-intermediate timescales, particularly for SPEI. Consequently, conditioning the log-log analysis on positive counts may omit part of the reporting response where drought primarily affects whether reporting is initiated, and comparisons among sectors, provinces, and timescales should consider both occurrence and conditional intensity.

A further source of uncertainty in the province-level spatial analysis concerns the independent extraction of impact categories and geographic references. In articles mentioning several provinces and several impact types, the available workflow cannot determine whether every reported impact applies to every referenced province. Because the final analytical dataset retains monthly province–sector counts rather than article-level provenance, the frequency of such combinations cannot be quantified retrospectively, nor can the analysis be restricted to single-province articles. This limitation is particularly relevant to the interpretation of fine-scale spatial contrasts. Accordingly, province-level coefficients and maps should be viewed as patterns of the geographic and sectoral framing of drought in media reporting, rather than as precisely geocoded inventories of physical impacts. Aggregated national and broad regional patterns are less dependent on exact within-article attribution, whereas interpretation of individual provinces warrants greater caution.

We report a dominant large-scale signal of ~~drought-impact reporting~~drought impacts, but with substantial ~~a large~~ spatial heterogeneity in sensitivity and dominant timescales. The first principal component indicates that a large part of the variability of ~~impact reporting~~the impacts is associated with ~~due to a~~ coherent climatic forcing over the whole of Spain. Several processes could plausibly contribute to the observed geography. Large-scale atmospheric circulation structures drought variability across Spain. Previous studies have linked drought variability in Spain to large-scale atmospheric circulation patterns (Vicente-Serrano and López-Moreno et al., 2006; Russo et al., 2015; Bakke et al., 2023; Noguera et al., 2024). However, basin storage, groundwater exchange, irrigation dependence, crop composition, water-allocation rules, and adaptive capacity can alter how the same climatic deficit is experienced and reported. Numerous studies demonstrate that local hydro-climatic conditions, water management practices and socio-economic factors modulate drought responses, with regional differences in sensitivity and dominant timescales (Iglesias et al., 2010; Grau-Satorras et al., 2016; Urquijo et al., 2015; Estrela et al., 2016; Iglesias et al., 2021; Li et al., 2024; Junquera et al., 2025). In addition, province-specific reporting intensity and the geographic focus of a national newspaper may contribute to apparent spatial contrasts. The spatial heterogeneity therefore likely combines hydro-climatic, management, exposure, and reporting effects and should not be attributed to climate alone. Crucially, ~~In the meantime~~, this spatial variability further demonstrates that drought impacts cannot be fully understood without considering both large-scale climatic forcing and local environmental conditions.

Dominant timescales are identified to offer further insight into the temporal organization of ~~drought-impact reporting~~drought impacts. Hydrological reporting is predominantly associated with medium- to long-term accumulation timescales, while agricultural reporting is more linked to short- to medium-term accumulation timescales (Wan et al., 2018). Both sectors exhibit a shift toward longer dominant timescales with increasing drought severity, indicating stronger associations with longer climatic accumulation windows. This change is particularly evident in hydrological systems, but it also emerges in agricultural systems under severe conditions, where agricultural reporting also becomes associated with longer accumulation timescales. Altieri et al. (2025) reported similar results indicating that even resilient agricultural systems can collapse under prolonged

water stress. The limited contribution of short-term timescales further supports the conclusion that media-reported impacts are more strongly associated with persistent, rather than persistent-not-transient, anomalies.

These findings are relevant for drought detection and the development of early warning systems. Impact dynamics are strongly controlled by drought accumulation ~~timescales hence~~timescales; hence multiscale monitoring strategies are required to achieve a complete picture of the evolution of drought. A number of studies have underscored the need to choose suitable drought timescales given by sector-specific responses to assess their severity (Vicente-Serrano et al., 2021). The relatively greater impact of atmospheric evaporative demand indicates the requirement for integrated temperature-sensitive indicators like the SPEI in operational monitoring networks. Although news articles are written for purposes other than systematically documenting drought impacts, they provide a credible, though not exhaustive, record of drought's effects at specific places and times (Dow, 2010). In contrast to SPI and SPEI, which quantify meteorological moisture deficits and atmospheric evaporative demand, media-reported impact data document how drought was observed, narrated, and experienced through effects on people, livelihoods, ecosystems, water resources, agriculture, and socio-economic systems. In this study, these records are therefore treated not as direct physical measurements of drought damage or as a comprehensive inventory of all impacts, but as a best-available proxy for the societal visibility, timing, location, and sectoral framing of drought events. ~~persistent~~Persistent conditions could generate repeated reports because no event-level de-duplication was applied. Thus, high reporting frequencies may partly reflect sustained media attention to the same ongoing drought consequence rather than a larger number of independent physical impact events. Because news coverage is shaped by editorial priorities, institutional context, public attention, event novelty, perceived severity, and uneven spatial coverage, low reporting frequencies should not necessarily be interpreted as the absence of drought impacts. Rather, media-derived impact data identify when and where drought became socially salient, problematic, or publicly visible, making them useful for relating climate-based drought indicators to observed societal consequences. Recent research has demonstrated that media-based data-sets can record drought effects at relatively high spatiotemporal resolution (Madruga de Brito et al., 2020; Wang et al., 2025), although they entail reporting biases and changes in media practices (Stahl et al., ~~2015~~2016; Serrano-Acebedo et al., 2026; Pita Costa et al., 2024).

The findings support the growing view that drought impacts in Spain emerge from the coupled interaction of climatic variability, hydrogeological structure, land management, and anthropogenic water use (Junquera et al., 2025; Sánchez and Cantos, 2025). Recent basin-scale analyses have shown that hydrological deficits across the Iberian Peninsula are strongly conditioned by groundwater exchange, irrigation pressure, permeability, and reservoir regulation, producing substantial spatial heterogeneity in effective water availability (e.g., Segura-Méndez et al., 2026). Within this context, the spatial variability of ~~drought-reporting drought-impact~~ sensitivity coefficients identified in this study likely reflects not only differences in meteorological drought severity but also the varying capacity of socio-hydrological systems to buffer, amplify, or redistribute drought stress. These factors provide plausible physical and management contexts for the regional differences observed here, but the media-based analysis cannot separate their individual contributions from differences in exposure or reporting.

655 Accordingly, spatial variation in sensitivity coefficient (β) can be interpreted as variation in the drought–reporting relationship rather than as a direct map of physical vulnerability.

This study advances understanding of drought-impact relationships by providing a comprehensive multiscale and nonlinear framework that links climatic drivers to societal responses under changing environmental conditions. The combination of
660 climate indices and impact data provides a good basis for further improvement of the assessment, attribution and prediction of drought impacts, thus facilitating more effective risk management and adaptation efforts in drought-prone regions.

5. Conclusions

This work provides a detailed assessment of the multiscale and nonlinear relationships between drought severity and media-reported hydrological and agricultural impacts in ~~peninsular~~ Spain during 1976–2023. We explored relationships ~~between of~~
665 standardized drought indices (SPI and SPEI) ~~and with impact-reporting impact~~ data from newspaper records. The results indicate that ~~the frequency of media reports on impacts media-reported impact frequency~~ is more strongly associated with persistent moisture deficits than with short-term climatic variability. The most robust ~~drought-reporting drought-impact~~ relationships are identified at medium-to-long accumulation time-scales, reflecting the integration of moisture deficits over longer climatic windows. These results indicate that the widespread impact reporting in ~~peninsular~~ Spain is more strongly
670 associated with persistent drought conditions than with isolated precipitation deficits.

The results also suggest a clear sectoral differentiation in dominant accumulation timescales. Hydrological reporting is mainly associated with longer accumulation timescales, following the integrative nature of water storage systems, whereas agricultural reporting shows stronger associations with short- to medium-term accumulation timescales, ~~reflecting the dependence of~~
675 ~~agricultural systems due to their dependence~~ on soil moisture and seasonal water availability. Despite these differences, both sectors exhibit a systematic increase in dominant timescales with increasing drought severity, suggesting that longer deficits are increasingly associated with reporting dynamics across both sectors.

A major finding of this study is the ~~the~~ strong association of atmospheric evaporative demand with ~~drought-impact~~
680 ~~reporting media-reported drought impacts~~. The consistently stronger relationships obtained with SPEI, compared to SPI, underline the importance of temperature-driven processes, especially for moderate to extreme drought conditions and at longer accumulation time-scales. The results ~~highlights~~ the increasing role of warming in drought dynamics and ~~suggests~~ that precipitation-based indices alone may underestimate drought severity and related impacts in a changing climate. The analysis
685 also indicates that reporting sensitivity to drought severity is largely nonlinear, increasing disproportionately for severe and extreme conditions. This behaviour reflects a threshold-like increase in reporting sensitivity and should not be interpreted as a direct measure of physical damage thresholds.

Complementary uncertainty and zero-inclusive analyses further show that these sensitivity patterns are more robust at the aggregated national scale than for individual province–timescale regressions, and that increasing drought magnitude affects both the probability that reporting occurs and the intensity of reporting conditional on its occurrence. This distinction strengthens the interpretation of the nonlinear drought–reporting response while emphasizing appropriate caution in the use of isolated local β estimates.

Spatial analysis indicates a dominant large-scale drought signal and ~~large~~ large regional variability in reporting response. Although a large proportion of reporting variability is associated with coherent climatic forcing over large areas of ~~peninsular~~ Spain, regional differences in sensitivity and dominant timescales suggest the importance of local hydro–climatic conditions, water management and socio-economic factors. These results highlight the importance of accounting for both regional characteristics and large-scale drivers in the assessment of drought impacts.

This study contributes to the understanding of drought-impact relationships by assembling a unified framework of multiscale drought metrics, nonlinear sensitivity analysis and impact-based observations. The results have important implications for drought monitoring and early warning systems, underscoring the importance of multiscale approaches and the inclusion of temperature-sensitive indices such as SPEI. The inclusion of media-reported impacts ~~provides useful insight into how drought manifests in society~~ ~~provides a useful insight into the way drought manifests itself in society~~, beyond the traditional climate-based indicators. This framework supports improved assessment, attribution, and prediction of drought impacts in the context of ongoing climate change and ultimately supports more effective risk management and adaptation strategies.

Data and code availability

The drought-impact data used in this study were derived from copyrighted newspaper content from El País. Owing to licensing and data-sharing restrictions associated with the source material, the underlying article-level dataset and derived data containing information extracted from these articles cannot be made publicly available. Access to the original newspaper content is governed by the terms and conditions of the data provider. The drought-index data used in the analysis are described and cited in Sect. 2.2. The R code used for data processing, statistical analyses, and figure generation is available from the corresponding author upon reasonable request.

Author contributions

A.M.E. and S.M.V. were responsible for the conceptualization and ~~development of~~ ~~proposing~~ the research idea. A.M.E., F.D.C., and A.-H. developed the methodology, carried out ~~the~~ analysis and visualization, and wrote the original draft of the

manuscript. S.B., J.V.T., A.C., M.-A., B.T., and M.F. assisted in data collection, contributed to manuscript editing, and literature enhancement, and contributed to project administration and coordination. A.M.E., and S.M.V., and F.D.C. contributed to draft writing. All authors contributed to the manuscript by reviewing and editing the original draft.

720 Competing interests

The authors have no relevant financial or non-financial interests to disclose.

Computing environment:

All statistical analyses were performed in R version 4.5.2 (R Core Team, 2025). Data preparation, filtering, grouping, and aggregation were carried out using the dplyr and tidyr packages. The empirical standardization of the monthly drought-impact reporting series was implemented following Vicente-Serrano et al. (2025) using custom R routines. Principal component analysis (PCA) was performed using functions from the stats package, separately for the standardized impact-reporting hydrological-and-agricultural-impact series. Pearson correlation coefficients between drought indices and standardized impact-reporting impact frequencies were also calculated using stats. Drought-reporting sensitivity coefficients Drought-impact sensitivity coefficients were estimated using ordinary least-squares log-log regression with lm(). Temporal dependence in the drought-index, impact-reporting, and regression-residual series was evaluated using the autocorrelation function (acf()) implemented in stats. Province-level spatial data were handled using the sf package.

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