



Christel Prudhomme¹, Anne K. Fleig², Reinhard Schiemann³, Christoph Frei³, Lena M. Tallaksen² & Hege Hisdal²

¹ Centre for Ecology and Hydrology, Wallingford, OX108BB, UK, chrp@ceh.ac.uk; ² Department of Geosciences, UiO, P.O.Box 1047, Blindern, 0316 Oslo, Norway, a.k.fleig@geo.uio.no
³ Federal Office of Meteorology and Climatology MeteoSwiss, Krähbühlstrasse 58, 8044 Zurich, Switzerland, reinhard.schiemann@meteoswiss.ch

Flood flow occurrence (CEH)

Hydrological drought (UiO)

Precipitation mapping (MeteoSwiss)

Aims and Objectives

Flood occurrence has been linked to weather types of the Grosswetterlagen classification (manual classification) in Germany [1]. This pan-European study evaluates (1) if such links exist anywhere in Western Europe, (2) if some weather types are associated to large-scale floods and (3) if results depend on the classification algorithm, using the objective classifications developed within COST733. The existence of significant relationships would show the hydrological relevance of the corresponding classification. This could be exploited by using the same algorithms to derive Weather Types (WT) from GCM outputs and anticipate large-scale floods as a seasonal time frame or evaluate flood probability at a future multi-decadal time horizon.

Data and methods

Daily flow series from over 400 catchments were obtained from the EWA, BRDC, NRFA & HYDRO archives. Flood events were identified following the Peak-over-Threshold method [2] (average of 3 peaks per year). Two hypothesis were tested and their significance level evaluated:

• **P1:** Is a weather type occurring more frequently during a flood event than usual?

$$PI1_{WT,season} = 100 * \left(\frac{n_{day_{season}} \text{ Flood with } WT_i / n_{day_{season}} \text{ Flood}}{n_{day_{season}} \text{ with } WT_i / n_{day_{season}}} - 1 \right)$$

• **P2:** Is the persistence of a weather type followed by a flood event?

$$PI2(i) = pr(WT = i \text{ for } \geq k \text{ days}, 0 \leq k \leq N^*)$$

P2 is compared with the binomial probability of at least k days out of N* of WT_i using historical frequencies of occurrence. Here N* = 10 days. For each station, a catalogue of flood event was constructed, the weather type associated with each event and up to N* preceding days identified and P1 and P2 calculated.

Results

For each indicator and time lag considered, results are displayed on maps showing, for a given WT, the associated indicator value (size of dots) and level of significance (colour). The Pan-European relevance of a weather type is assessed by histograms of the percentage of stations falling into a certain bracket value.

Figure 1 illustrates P1 for winter for two WTs of the objective Hess Brezowski classification (OGWL) derived from ERA-40 re-analyses developed within the COST733 [3]. For the great majority of stations analysed, WT2 occurs in winter more often 3 days before a flood event than on any other days (a). At the opposite, there is no significant association between WT9 and winter flood occurrence (b)

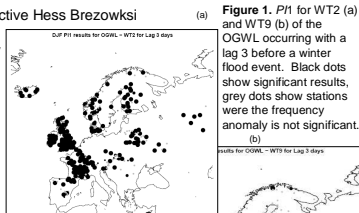
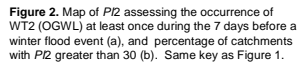


Figure 2 shows that all over Europe (a) WT2 occurred at least once in a window of 7 days before a winter flood event and significantly more often than would be expected by chance (PI2 with N* = 7 days, k=1). Over 60% of the tested sites have a conditional probability P2 greater than 30 : WT2 occurred more than 30% of the time before a flood event (b), more often in this season than would be expected by chance.



Conclusion

The results presented were obtained with OGWL and investigated possible association of some Weather Types with the occurrence of large floods in Europe. Evidence was found that, in winter, WT2 occurs more often before a flood than the rest of the time and when a flood occurred, more than 30% of the time WT2 occurred at least once in the 7 days preceding the flood, more often than expected from chance alone. The increase of WT2 occurrence in the future might thus be associated with higher flood risk in Europe. Further research is needed to test other classifications, and to evaluate if at the opposite, some WT never precede a flood (e.g. WT9), which could suggest lower flooding risk if their occurrence increases.

References

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Global Runoff Data Centre (http://grdc.bafg.de/services/Entry_987_Display)
European Water Archive (<http://ewa.bafg.de>)
UK National River Flow Archive (<http://www.ceh.ac.uk/data/nrfa/index.html>)
Banque Hydro, France (<http://www.hydro.eaufrance.fr/>)

Aims and Objective

In addition to flood, drought is the other hydrological extreme that can cause severe problems. Droughts are slowly developing and severity increases with increasing duration and extend. By identifying weather types (WT) which contribute to the development of droughts, hydroclimatology processes leading to severe hydrological droughts can be studied. The objective of this study is to compare objective weather type classifications (WTCs), with respect to analyzing links between weather types and hydrological drought in north-western Europe. The inter-comparison considers (1) classification algorithm, (2) input variables and (3) number of defined WTs.

Data

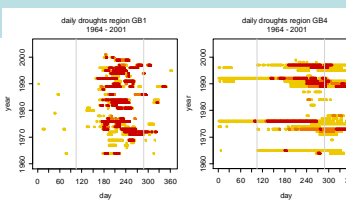
Regional hydrological drought: Daily Regional Drought Area Index series (RDAI; 1964–2001) for four regions in Great Britain and two in Denmark [1].

RDAI: - based on at-site drought series derived from streamflow records [2];
- represents the proportion of the total drought-affected catchment area within a region.

Drought: - defined as RDAI > 0.7.

Regional hydrological drought characteristics: vary between the regions, e.g.
- short but frequent droughts (Scotland, GB1);
- few but long droughts (southern and central England, GB4; Figure 1).

Figure 1 Regional drought development in regions GB1 and GB4; (RDAI; 1964–2001) for four regions in Great Britain and two in Denmark [1]. Orange: 0.5 < RDAI < 0.7; Yellow: 0 < RDAI < 0.5.



Weather Type Classifications: Daily catalogues of the 71 automatic WTCs from COST733 [3] and the subjective Hess-Brezowski Grosswetterlagen (HBGWL [4]) are used: - 24 different classification algorithms;
- one or two sets of input variables (SLP as common one);
- different numbers of WTs (approx. 9, 18 and 27).

Method

Identification of WTs, which may be associated with the development of hydrological drought (for each region):

I: Frequencies of WTs prior to and during the five most severe droughts events ($F_{e,WT}$) are compared to the normal frequencies of the weather type during the same period of the year for 1961–2001 (F_{WT}).

$$- \text{WTs with } \mu F_{e,WT} = \frac{1}{5} \sum_{e=1}^5 \frac{F_{e,WT} - \mu F_{e,WT}}{\sigma F_{e,WT}} > 0 \text{ are selected as group } WT_{pos}$$

- Duration of the considered period depends on the regional drought characteristics. ($d_{region} = 30$ to 180 days).

II: Correlation analyses for the summer period only (16 April – 15 October) between WT_{pos} frequencies (total of all WTs in WT_{pos}) and drought to compare WTCs:
- daily: moving d_{region} -day sums of WT_{pos} frequencies; and daily RDAI;
- seasonal: total summer WT_{pos} frequencies; and number of drought days during the summer (for GB4 WT_{pos} frequencies during the summer + the previous winter are used).

Results

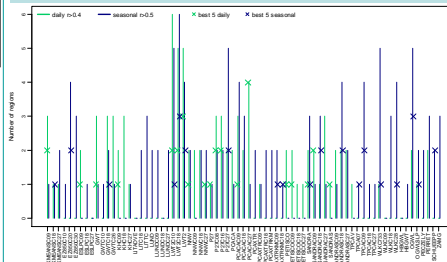


Figure 2 Number of regions for which a WTC (1) obtained a correlation coefficient r > 0.4 and 0.5 for the daily (green bars) and seasonal (blue bars) analysis, respectively. (2) is among the five WTCs with highest r-values (crosses).

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Motivation

Gridded precipitation climatologies based on high-density station networks and spanning several decades are available in the Alpine region (e.g., Figure 1, [1]).

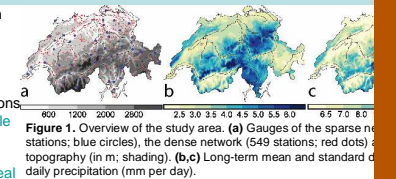
• However, there is a time delay (typically days-weeks) until observations from all stations are digitised and much less data are available for quasi real-time precipitation gridding.

• How to approach this problem?
Can weather-type information help in near-real time gridding of daily precipitation?

Here,

• We test a reduced-space optimal interpolation (RSOI) method for the construction of daily precipitation from a sparse gauge network.

• Stratify the method according to a weather types classification [2,3] and compare with results from the unstratified interpolation.



Reduced-space optimal interpolation

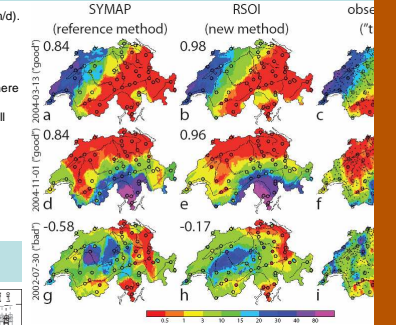
• is an interpolation method normally applied in climate reconstruction. Here, it is tested for near real-time gridding of daily precipitation.

• RSOI combines information from

(i) high-quality precipitation grids based on the dense network (for calibration, not available in real-time) (ii) sparse gauge data available in quasi real-time
• RSOI is based on principal component analysis of the calibration data, truncation of the data space, estimation of principal component scores from the sparse gauge data. See [4] for details.

Examples

Figure 2. Examples of gridded daily precipitation (mm/d). (left) Gridding from the sparse network in terms of a reference method [1], (centre) RSOI from the sparse network, (right) precipitation grid based on the dense network. The top two rows show cases with good interpolation results, the bottom row shows a case where gridding from the sparse network is very difficult. Numbers show the value of a mean-squared-error skill score used to evaluate the interpolations.



Weather types

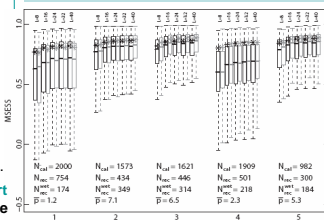


Figure 3. Boxplot statistics for the distributions of daily mean-squared-error skill scores for unstratified RSOI (black) and RSOI stratified with respect to the five PCA(2,3) weather types (grey). For each weather type, the annotations are (i) the number of days in the calibration period, (ii) the number of days in the reconstruction period, (iii) the number of wet days (with domain-mean precip > 1 mm/d) considered in the evaluation, and (iv) the domain-mean precipitation for the weather type considered. Results are shown for different values of the reduced-space dimension L (a parameter of RSOI).

Conclusion

• Reduced-space optimal interpolation is suitable for gridding daily precipitation data.
• For the setup tested here, RSOI clearly outperforms the reference method (based on climatological data) the SYMAP algorithm, [1]).
• Stratification with respect to weather types clearly improves the interpolations for some weather types.

References

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