Interactive comment on “Positive and negative human-modified droughts: a quantitative approach illustrated with two Iranian catchments” by Elham Kakaei et al.

Elham Kakaei et al.
hrmoradi5@gmail.com

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The paper investigates human-modified droughts; an interesting topic in the scope of the journal and at least theoretically well supported by the background literature in the introduction. The analysis is based on a framework recently proposed by van Loon et al. (2016a,b) to distinguish between climate- and human-induced and human-modified drought events and to disentangle the causes for drought, namely climate variability and anthropogenic forcing due to water abstractions from rivers and groundwater. The method differentiates between periods of natural and disturbed streamflow and utilized then a N-A-model (calibrated during the natural period) to calculate natural streamflow during the proposed period of disturbed streamflow. In principle the paper proposes that different drivers of streamflow droughts can be counted against each other to gain a cumulative drought signal. The authors concluded that for the two presented catchments human activities have caused mostly negative human-modified droughts. Although the research topic is important I doubt that structure of the methodology, the presentation of the results and their interpretation lead to clear conclusions and advanced implications. I missed a clear and structured development and description of the method. I think the editor has to decide whether major revisions and structural and graphical improvements are needed or a new submission of the paper is the better way. I encourage the authors to revise the method description, the analysis and the graphs to gain a more comprehensive paper.

- The authors appreciate the reviewer’s constructive comments. The comments have been incorporated in the revised manuscript. Responses to the specific comments can be found below. We trust that the revisions we propose support better the conclusions and the implications. We think that the stepwise approach from Step 1 to Step 5 (Sections 2.2-2.5), which is presented in Figure 1, provides a well-elaborated description of the methodology.

Main comments

+ STUDY CONCEPT As the study concept (components of droughts) is rather new I suggest to explain the different types in more detail. Why were exactly these two catchments chosen for the analysis and what are the “major concern” (p2132) in detail regarding the human-modified droughts here? Section 2.2.1 and 2.2.2 could be shorten (well-known statistical tests), but more justification is needed to proof that e.g.
Pettitt’s Test is really appropriate to distinguish periods of natural and disturbed streamflow (e.g., change points can also emerge due to climate change). Also the explanation of the anomaly analysis isn’t straightforward for me; what is surplus value of all these equations and subscripts (eq. 8-11) if the concept could be explained with a good overview figure?

Response:
- The different types of drought, that is, climate-induced drought, human-induced drought and human modified drought are explained in more detail in Section 2.5 (P8, l12-31 and P9, l1-19).
- The Eskandari and Kiakola catchments in Iran have been chosen, because we would like to have catchment from a dry climate that face climate-induced drought and usually have to cope with human interventions that cause the drought characteristics to change, either in a positive or in a negative way, i.e. positive human-modified drought or negative human-modified drought, respectively. The drawback is that catchments from dry environments often are data sparse. We anticipated that this is acceptable because the focus of the paper is to propose a methodology to break down the human-modified droughts into positive and negative ones. We have added a line in the revised manuscript to explain this (P2, l30-33, P3, 127-129).
- The main concerns about human intervention in both Iranian catchments are land use change, incl. crop pattern change, large groundwater abstraction and population growth. We have explained these human interventions in Annex 1 to this reply. We suggest not to include all these details in the main text of the paper, because the focus of the paper is on the methodology, as said above. Alternatively, we can add the text as an Annex to the paper.
- We agree with the reviewer that the description of the well-known statistical tests can be shortened. Hence, Sections 2.2.1 and 2.2.2 are made briefer, and furthermore a more-elaborated justification is presented to proof Pettit’s test ability to distinguish between natural and disturbed periods. The results of Pettit’s test for rainfall time series in both catchment are added to Annex 2 of the paper (P37, Table 1) and (P38, Figure 2). Furthermore FDCs of monthly observed discharge records were plotted for the natural and disturbed periods (P39, Figure 3). As explained in Section 3.1 (P10 l22-l27 and P11 l1-l7), in the Eskandari catchment, the change point in discharge and rainfall time series approximately occurred in May 1996 and October 2001, respectively. In addition, according to the Mann-Kendall test (Section 3.1, P10 l5-l22), the discharge showed a clear significant downward trend, while, rainfall showed a weak none-significant upward trend, which indicate the influence of both natural (weather variables) and human drivers on discharge. Based on the Pettit’s test the period 1976-1996 could be classified as the natural period (undisturbed period) and the period 1996-2014 as disturbed. In the Kiakola catchment, the change point in discharge and rainfall time series roughly occurred in May 1998 and September 2004, respectively. Considering the result of Pettit’s test on discharge and rainfall and the Mann-Kendall test results in Section 3.1 (P10 l5-l22) (significant downward trend in discharge, significant upward trend in rainfall, and a none-significant upward trend in evapotranspiration), the periods 1976-1998 and 1998-2013 could be considered as natural and disturbed, respectively. For both catchments (Eskandari and Kiakola), the change point in rainfall time series occurred 5 years (2001) and 6 years later (2006) than changes in discharge (1996 and 1998, respectively). So, the change point in discharge has mainly to be contributed to human intervention in both catchments. We would like to stress here that the application of statistical tests is important in the overall methodology to distinguish between the natural and disturbed periods (Step 1). The actual results for the two Iranian catchment are less important, as mentioned above. These only illustrate what need to be achieved in Step 1.
- The equations of climate-induced drought, human-induced drought and human-modified drought formalize the definitions of each type of human-modified drought and complement the conceptual diagram Figure 3. We believe that these equations have added value and leave no room for different interpretation. The equations expand on those presented by Tallaksen et al. (2009). We added some text to clarify this (P8

C3
+ HBV MODELING For me it is not possible to evaluate whether the model is appropriate to answer the research questions or not. For example Fig.6 shows huge discrepancies in observed and simulated streamflow (btw: Why is the black line here needed if the observed discharge is only compared to the threshold?). The authors stated that the model performance is appropriate, but in my eyes the model performance is rather poor suggesting that the model setup/ preparation/ calibration has some issues. Is NSE or logNSE used (it’s not clear from the text, p7l6)? However, is NSE really the best OF or would be a measure with more emphasis on volume errors a better choice? How valuable are the results (Tab.2+3) showing mean drought deficits of 0.7 mm to 4.1 mm if they can easily attributed to model simulation uncertainty? The different thresholds in Tab.5 somehow shows that higher thresholds (50th) are needed to gain significant deficit volumes in respect to drought management. If so, I ask myself if a 50th threshold is really valuable for drought assessment.

Response:
- The natural time series is shown in Figure 6 (P31) to reflect natural conditions without human drivers of drought. The natural time series are an essential component of identifying human-modified droughts (Van Loon et al., 2016ab; Van Loon and Van Lanen, 2012).
- The model performance is evaluated based on the lnNSE, which is an appropriate objective function for low flow simulation (P7 l9-I21). It has been frequently used. Each of the statistical criteria has specific pros and cons, which have to be taken into account during model calibration and evaluation. The Nash-Sutcliffe efficiency (NSE) and the coefficient of determination are very sensitive to peak flows, at the expense of being less sensitive to low flow conditions. They are based on the squared differences between observed and simulated values (Pushpalatha et. al., 2012; Krause et. al., 2005).

Additionally, the coefficient of determination alone should not be used for model evaluation, because it can have still high values for very poor model results, because it is based on the correlation only. In order to reduce the problem of the squared differences and the resulting sensitivity to high extreme values, the Nash-Sutcliff efficiency is often calculated with logarithmic values of observed and simulated values (lnNSE), in particular if drought is considered. Through the logarithmic transformation of the discharge values the peaks are flattened and low flows are kept more or less at the same level. As a result the influence of the low flow values is increased in comparison to the peaks values resulting in an increase in sensitivity of lnNSE to systematic model over- or underestimation. So, the lnNSE reacts less on peak flows and stronger on low flows than the NSE (Krause et. al., 2005). So, the logarithmic Nash-Sutcliffe efficiency according to the algorithm of the observed and simulated flow discharge (lnNSE), as the best objective function for low-flow modelling, was utilized for the assessment of discrepancies between observed and simulated discharge and the performance of the HBV in both catchment
- The NSE≥0.5 has been defined as an acceptable value for model performance (Christiansen, 2012; Moriasi et. al., 2007). Moreover, we would like to stress again that the focus of the paper is on the methodology rather than on the actual outcome for the Eskandari and the Kiakola catchments. Hence, model performance is not of utmost importance (P7, l22-l24), as long as it passes the minimum criterion.
- Mean drought deficits are small, because minor droughts with small deficits are included as well, although events <15 days were excluded from the analysis. Clearly, model uncertainty will affect the numbers, but outcome is still valuable because other model structures will result in similar relative small numbers. As said before, the focus of the paper is on the methodology rather than on the exact numbers for the two Iranian catchments. These are only in the study to demonstrate the methodology.
- Selection of the threshold level is quiet challenging. For drought management, not only the yearly recurring (summer or winter) low-flow period is important, but any deviation from the normal seasonal pattern (Van Loon and Van Lanen, 2012). Most of
the time in arid and semi-arid regions the discharge is zero or close to zero and the
time series of the data sets (discharge and rainfall) contain a significant number of ze-
ros values. In the Iranian study regions, there is a strong seasonality in rainfall and
discharge regimes and it is not uncommon to find extended dry periods up to several
months where no rainfall is observed, i.e. zero values (Favier et al., 2009; Verbist et al.,
2010; Meza, 2013). The threshold of 50th percentile could have been used to identify
drought events to avoid a threshold of zero (Rangecroft et al., 2016; Giannikopoulou et
al., 2014; Van Huijgevoort et al., 2012). We tested the effect of using different threshold
values on drought characteristics (Section 4), and eventually on the percentage of neg-
ative and positive human-modified droughts (Step 5) by using the 50th, 70th and 90th
percentiles. As hypothesized, choosing different percentiles as threshold will change
the magnitude of the drought characteristics. A lower threshold (90th percentile) leads
to fewer human-modified events with shorter durations and lower deficits, whereas,
higher thresholds (50th and 70th percentiles) results in the opposite. Eventually, the
magnitude of the threshold has to be defined by the drought manager or stakeholders.
For perennial streams thresholds of 70-90th are commonly used. In semi-arid and arid
climates (e.g. Iran), higher thresholds, e.g. 50th, could also be used and provide rele-
vant information (Fleig et al., 2006). In the paper we have selected the 80th percentile,
because it has been used in most drought studies, but we also included higher and
lower thresholds. This has not been done to discuss drought management (not the
aim of the paper), but because we would like to show that the selection of the threshold
does not affect the pattern of negative and positive human-modified drought (which is
the purpose of the paper).

+ DATA I understand that these kind of studies in data-scare and semi-arid regions are
very important for local water management. However, in this case a relatively novel
method is used to investigate different drought types. Would it be better to have a
larger (and perhaps better) data set to eliminate all the catchment-specific issues of
the two proposed catchments? At least the method could be conducted on a larger
data set including the two Iran catchments. However, if it is really important to under-
stand climate- and human-influence in these catchment, then a clear justification for
the change points in the time series have to be made (e.g., due to better visualizations
of the change).

Response: - The paper proposes a step-wise methodology to expand on the distinction
between different hydrological drought types in the Anthropocene, in particular making
a distinction between positive human-modified and negative human-modified droughts.
The proposed methodology, as a multi-directional and multi-driver drought framework,
enables a further elaboration of drought in the Anthropocene. So, it makes it possible
for water resources managers to decide on how to combat natural and human-made
droughts. Long time series of hydrometeorological data for both natural and disturbed
period are the basic elements of the proposed methodology, irrespective of the region.
We decided to use data from a dry environment, because climate-induced droughts are
frequent there and human interventions are common, likely causing human-modified
drought. We realized that many of these time series have quality issues (e.g. short
time series, gaps). In addition, this kind of information is not available because of water
conflicts and security reasons in many arid and semi-arid regions. However, we think
that the data of the two Iranian catchments are sufficient to illustrate the potential of the
proposed methodology. In a follow-up study a larger set of catchments can be analyzed
to explore differences between different climate settings and human influences. The
paper certainly has not the aim to understand in depth the climate and human influence
on drought types in the two Iranian catchment.

+ GRAPHS The quality and in-depth information of the single graph is often below aver-
age. Fig.2 is hardly readable due to graph quality issues. For me it is not possible with
Fig 4. to see how good/or bad the model is performing, the lines are too tick, the axis
labels are not appropriate (here more in-depth analysis is needed, e.g. FDCs), Fig.5-7
show examples but no systematic visualization of different drought types, Fig.1 (An-
nex) try to highlight the change point in the time series, but the reader cannot find any
justification for this statement (improvement: remove the points, check if y-log-scale
would be helpful, compare annual streamflow series from the period before/after the proposed change point. Overall some synthesis graphs are really missing to support the points the authors want to make.

Response:

- We revised a number of graphs. Figure 2 has been revised and hence the quality improved (P27). Figure 4 is replaced by the FDCs of observed and simulated discharge for the natural period (P29). Figures 5-7 are revised (P30-32). Figure 1 (Annex 2) has been deleted and the change point in the discharge and rainfall time series are put on a y-logarithmic scale (P37 Figure 1 and P38 Figure 2). In addition, the FDC of monthly discharge has been added for the natural and disturbed periods before and after the change point (P39 Figure 3).

- We think that the synthetic graph (Figure 3) together with the equations adequately present the generic points we would like to make.

References


Annex 1: Human interventions in the Eskandari and Kiakola catchments

During the past decades water resources management in the Zayandehrood river basin, which includes the Eskandari as an important tributary, has been a critical issue, because of the growing human population, industry, high agricultural water use and frequent droughts, land use change, incl. decrease of pasture land and forests (Mirzaei et. al., 2016; Gohari et. al. 2017). These have caused unprecedented water shortages in the basin. Meteorological and hydrological extreme events are common in this basin (Madani and Mariño, 2009). The Zayandehrood River, as the backbone of human development in central Iran, dries up seasonally, which imposes extensive pressure on the urban population, agriculture and industries (Madani Larijani 2005). Droughts in past decades has decreased the capacity of Zayandehrood dam’s lake to less than 150 MCM (Shafiee and Safamehr, 2011) and had harmful impacts on the water resources system of the area, such as reduction of agricultural and industrial water availability and dryness of the Gavkhooi swamp. The Gavkhooi swamp, as one of the most valuable ecosystems in Iran, is an important feature of sustainable development in Central Iran. The swamp provides habitat for over 140 bird species and numerous other flora and fauna. In addition, the Gavkhooi's ecosystem has a vital role in controlling water quality and stabilization of sand dunes located around the wetland. Over the years, population growth and inappropriate water resources management have resulted in a decrease of the quantity and deterioration of the quality of the wetland’s incoming fresh water and initiated destruction of this ecosystem (Sarhadi and Soltani, 2013). Some have little hope that this swamp will recover (Madani, 2014; Gohari et.al. 2014). The Eskandari catchment, as one of the most important tributaries of the Zayandehrood River upstream of Zayandehrood's dam basin, plays an important role in the possible restoration and preservation of
the river and Gavkhooni swamp. Inflow fluctuation due to frequent drought periods has recently become an important concern, also in the irrigation sector. The water shortage during dry seasons has intensified the effects of the inflow fluctuation on water users, especially in the downstream section (Hashemy Shahdany and Firoozfar, 2017). The residential areas have spread over the study area from 1997 onwards. Degradation of pasture land and their conversion into residential areas or bare lands over time has increased the surface runoff (Mirzaei et. al. 2016) leading to lower flow in the river during dry periods. The Department of Environment of the Mazandaran province, in the north of Iran where the Kiakola is located, has reported that there are too many wetlands, which have been exposed to drought due to climate change and anthropogenic activities (land use change and overexploiting of (sub-)surface water resources). In the past two decades, the forest area has decreased and the Kiakola catchment has been one of the most affected areas due to rapid urbanization and population growth (in the Mazandaran province, the number of people increased from 960,568 in 1986 to 2,010,948 people in 2011, Statistics Centre of Iran, 2011). This led to higher food and fuel demands, timber smuggling, more intensive livestock grazing, extra road development, exploitation of mines and construction of factories. Changing forest to grassland and overgrazing these grasslands by livestock, has become one of the most major concerns in this area. Grassland were also converted into residential areas. Comparing land use maps from the years 2000 and 2011 showed that rangeland was converted to residential sites (Zare et. al. 2017). A large area of forest land has been turned into gardens and crop land, including rice cultivation (Kavian et. al. 2018). Rice is the second highest consumption product in Iran and the demand is growing each year. Drought in different stages of rice cultivation impacted production (Hasheminya and Dehghannya, 2013).

Please also note the supplement to this comment: