

Interactive comment on “Climate change alters low flows in Europe under a 1.5, 2, and 3 degree global warming” by Andreas Marx et al.

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We thank the reviewer for the time and effort in commenting on our manuscript. We provide responses to each individual point below. For clarity, comments are given in normal font, and our responses are given as blue text. For newly produced figure 1, please refer to AC3 under reviewer 2.

This manuscript deals with a multi-GCM and multi-hydrological models assessment of changes in low flows across Europe between a present-day period (1971-200) and 3 different global warming levels: 1.5K, 2K, and 3K (and between them as well). It therefore contributes to document the effects of climate change on low-flow hydrology in Europe in the context of the Paris Agreement. This manuscript thus deals with a topical and important topic, and fits well into the scope of HESS. It is generally well structured and written, and conclusions are generally well supported by results shown. I have however two main comments (as well as specific comments) detailed below that should be addressed before the manuscript is published in HESS.

Thank you very much for the extensive commenting and feedback which will help to increase the quality of the manuscript.

1 Main comments

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1.1 Hydrological calibration and simulation over influenced catchments The calibration details (specific comment #9 and #10) as well as the validation results (specific comments #12, #13, #14) do not give enough confidence on the quality of hydrological modeling, and highlights the issue of calibrating and/or validating – seemingly natural-catchment-only – models against highly influenced catchments like the Ebro or the Rhône, especially for low flows. First there is not enough information on the calibration process, and even the catchments used for that are not identified. Second, validation is done for a large part over influenced catchments, and also over ensembles of highly nested catchments. Both points should be reconsidered in a future revision of the manuscript.

We acknowledge these facts. The information given on calibration and validation will be modified and extended in the manuscript. More information is given in the specific comments.

1.2 Scale of catchments selected for presenting and averaging results There are numerous inconsistencies throughout the manuscript in terms of the minimum catchment size used for presenting results (and giving averaged figures), see specific comments #20, #21, #22, #24, #27. Addressing this comment may imply reformatting all results, but this is also intrinsically linked to main comment #1. Indeed, the manuscript state that the runoff routing scheme prevent using results for catchments smaller than 10000 km², and near-natural catchments are usually only smaller than that in Europe. In parallel, maps of results are given over a river network encompassing drained areas much smaller than the indicated threshold. This thus shades doubts (maybe unjustified, but it has to be demonstrated) on the validity of models and results, together with issues highlighted in main comment #1.

The information on catchment sizes will be extended in the manuscript. Smaller catchments <10000 km² have not been omitted. The catchment size at a spatial resolution of 5 × 5 km² is limited e.g. by the DEM in determining the catchment boundaries. Therefore, results from catchments (or better river grid cells) with a contributing area >1000 km² have been used in the study, and these are shown in

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figure 3 and figure 5 and have been used for drawing conclusions. The selection of catchments >10000 km² in figure 1 and 4 has been done for clarity reasons (clearness of the figures) only. This will be changed for validation figure 1. More information is given in the specific comments.

2 Specific comments

1. P1L2, “1.5, 2 and 3 K”: please specify that this is with respect to the preindustrial period

Agreed.

2. P1L10, “-12%”: What is the baseline period here? This is all the more important that there could easily be confusion with the baseline used for the global warming level (see above).

Agreed. Baseline period for determining relative changes is 1971-2000. This information will be included.

3. L11-12: this sentence is ambiguous. Less snowmelt may imply less streamflow in some conditions (e.g. constant liquid precipitation or declining total precipitation). Please rephrase and make it clearer.

Agreed. We will modify the text as suggested.

4. P1L13-14: This sentence is also quite ambiguous. What is exactly preventing distinguishing between 1.5 and 2K warming effects? Is it the interannual variability which prevents distinguishing statistically estimates of period-averaged changes- for a given GCM-HM combination? Or is it the uncertainty due to the multimodel ensemble that prevents distinguishing ensembles of multimodel period-average estimates between present and future? Or both? Please make it clear here.

Agreed. It is both and we will include the information on uncertainty.

5. P2L11: The low-flow component of the 2015 drought event has been specifically studied by Laaha et al. (2017). I believe this reference is worth adding to the

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manuscript.

Agreed.

6. P2L21-27: What is the time slice that corresponds to the quantitative and qualitative results recalled here? Please make it clearer.

1971-200. We suggest to phrase "until the end of the century."

7. P4L7-10: First, this interpolation step should not be called downscaling as the latter refers to methods that actually add information for each day (either through regional climate models or empirical-statistical downscaling models) to the larger scale GCM fields. I would therefore strongly recommend using "disaggregating" or "disaggregated" instead of "downscaling" or "downscaled" in the manuscript.

We will modify the text as suggested.

8. P4L7-10: Second, this interpolation step should be better documented here, in order for the reader to understand the advantages and shortcomings of this approach, which are essential for assessing the quality of subsequent hydrological simulations. This interpolation step should ideally be assessed using a global reanalysis against high-resolution gridded datasets, like RCMs are actually assessed (see e.g. Kotlarski et al., 2017, among many others, for a recent example). This would critically allow distinguishing errors coming from (1) the spatial interpolation technique (and their large-scale forcings), and (2) the hydrological models. Please at least add some comments on that in the manuscript. Plus, the reference used for this interpolation technique is incomplete in the list of references.

We will extend the text as suggested. Assessing the meteorological input fields against other sources is out of the scope of this study. The missing reference will be added.

9. P4L24: What are these 9 catchments? Please provide some more information (location, surface, etc.). Are they near-natural or influenced catchments?

#9, #10 and #11 are commented together under #11

10. P4L24-26: What is the period used for calibration? And what are the calibration criteria (for both automatic and manual calibration)? Are they specific for low flows? Please carefully specify all this in the manuscript.

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#9, #10 and #11 are commented together under #11

11. P4L27-29, "The assessment ... (Gosling et al., 2017)": This is a very strong statement, which I tend to disagree with at least as a general conclusion. This is moreover hardly supported by the reference given in the manuscript, which compares global hydrological models and catchment hydrological models for the Rhine and Tagus (and other catchments, but not located in Europe). Results for a low flow indicator (Q95) show a large divergence of the two types of models with increasing global warming level (Gosling et al., 2017, their Fig. 2). As a conclusion, I would therefore strongly recommend removing this statement from the manuscript.

Comments #9, #10 and #11 are interrelated and are answered connectedly.

It is important to recognise that all HMs applied are well-established, widely applied and have been used (and calibrated) for Europe in former studies referred to in the manuscript. Furthermore, additional calibration for the three HMs was done in focus basins. Nevertheless, the validity of calibrated parameters may be limited in CC studies (?) and the results in multi-HM climate impact studies may be less influenced by the calibration than by the model-structure of the HMs. This could be shown e.g. in ?.

We agree to #16 that the statement "The assessment ... is independent" is too strong, we would remove this statement with the citation Gosling 2017 and replace it with "Well-calibrated HMs do not necessarily mean that future discharge under a changed climate can be reproduced satisfactorily (?). Furthermore, the selection of HMs may have a larger effects than calibration in hydrological climate impact studies (?)".

Considering the HMs calibration (#9 and #10) we suggest to extend the manuscript and include the paragraph: "Furthermore, the three HMs used in this study were calibrated in nine near-natural European focus basins located in Spain, Norway and UK, which were selected based on the consultation with the user groups within the EDgE project. Besides these, we also included three more central EU catchments (located in France and Germany) to represent diversity in hydro-climatic regimes. All HMs parameters were calibrated such that the model simulations represent a range

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of hydrologic regimes, rather than tailored to any specific characteristics. This was done in a consistent manner so that the model simulations can be used for a range of indicators (including high, low, and average flows) within the EDgE project. HMs could be calibrated to specific parts of the flow duration curve (FDC), however, this was not done in this study to avoid too specific tuning of the model simulations to those unique conditions and thereby losing valuable information on the entire FDC.

In the current simulations human water management was not taken into account, since some models lack the ability to include these processes and one focus on this work is on determining the HM uncertainty in low flow conditions. Human water management can however have a significance impact on the low flow conditions, due to abstraction of additional water in drought conditions or changes in reservoir management - as a result constraining the model to any specific low flow characteristic can result in a biased simulations. Also due to the similar reason we may expect a relatively lower model skill in matching the observed low flow characteristic."

12. P4L33-P5L4 and Figure 1: The assessment of HMs is very light and not strongly supported by Fig. 1. Indeed, this figure is potentially misleading, as it basically only checks that catchments have equally small/large indicators (Q90 or Q50) for both observations and simulations, which is mainly driven by the size of the catchment. I would therefore recommend using a different and more informative representation of differences, preferably in terms of relative errors (in percents), and also preferably as maps in order to show the potential spatial pattern in errors. This representation would also greatly help in comparing present-day errors to relative changes presented later in the manuscript. I personally would not give too much credit for a model showing for a given location present-day errors as large as 3K future changes...

We agree to change the metric to remove the catchment size effect. Therefore, we would use specific discharge [mm/d], include the information on the HM ensemble mean, and show the relative bias spatially distributed (see attached figure). We tend to disagree with the statement " I personally would not give too much credit for a model showing for a given location present-day errors as large as 3K future changes...".

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The comparison shown here is an "honest" one because the HMs are driven with GCM input for a time period in the past. This is usually not shown in climate impact studies. Furthermore, assuming a constant error or bias over time in the GCM-HM simulations would result in perfect study results. Therefore, we trust the uncertainty measures presented in this study (SNR combined with robustness) more. Considering the relative biases shown in the attached figure it is important to notice that the spatial pattern is very different from the climate change signal (Fig. 3 and 5). It would be critical if these patterns would match.

13. Figure2, right: This figure shows the location of validation gauges used in Fig. 1. First, it shows that many points in Fig. 1 comes from the same rivers and are necessarily highly correlated, which inherently bring some bias to the results that should be representative of the whole Europe. I would strongly recommend removing redundant points scatter plots like presented in Fig. 1. This would not be a problem however with suggested spatial representations (cf. above).

The validation gauges have been changed and more gauges are included in the revised manuscript. This will be changed (attached figure), and additionally, spatial representations are shown.

14. Figure 2, right: The second point is that several validation gauges are located on highly influenced rivers. For example, the Ebro river (Spain) is heavily influenced by water abstractions for irrigation, and the seasonal regime of the Rhône river (France) is heavily influenced by all the hydropower reservoirs located in the Alps (and other surrounding mountain ranges). There are many other cases that can be spotted on the map. As a consequence, observed streamflow indicators for low flows simply cannot be compared to natural (i.e. without human influence) hydrological simulations for these catchments. A good fit to observations may indeed reveal that physical parameters in HMs are tweaked to compensate for no representation of human influence. This may not be a problem in itself (at least for practical modeling purposes if not scientifically satisfactory) if human influences would not have changed and would not change in the future. Which has happened and definitely will. As a conclusion,

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I would strongly recommend using only near-natural catchments as validation (and also calibration) gauges for natural hydrological modeling (as I suppose it is the case in the manuscript, even if some HMs considered may represent human influences). A number of reference hydrometric networks have recently been developed at the country scale (Hannaford and Marsh, 2008; Giuntoli et al., 2013; Murphy et al., 2013), and one should take advantage of these. Note that these networks overlap for some countries (but not for some other) with stations tagged “climate sensitive” in the Global Runoff data Centre.

Calibration in HMs has been performed using headwater catchments and no heavily human influenced basin was included. It would generally be a good idea to use “climate sensitive” stations only. Unluckily, these are not uniformly distributed all over Europe, but only available in selected countries. Esp. in the Mediterranean area there is no such station available. We would consider this comment in future studies in case an area-wide coverage of climate sensitive stations is available.

15. P6L3: The 0.46K figure has uncertainties attached to it, according to the reference cited (Vautard et al., 2014). Please do mention these uncertainties in the manuscript, with possibly additional references that provides 1971-2000 estimates of global warming level.

Agreed. "The warming of 0.46 K in an average value from three estimations with a spread between 0.437 K and 0.477 K."

16. P6L20: The use of calendar year is not entirely satisfactory for computing Q90 in snow-influenced catchments where the low-flow period (or one of the low-flow periods, which is a more difficult situation) may span two calendar years. Please consider changing the calculation procedure or at least justify this approximation.

We will extend the text including the limitation mentioned.

17. P7L8-9: Please mention here (rather than in the results section) that the robustness is compute as the percentage of projections showing a significant change.

Agreed.

18. Table 2: Please make clear that “1980s” refers to the 1971-2000 period.

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Agreed.

19. P8L3-9: I don't really understand this peculiar choice of method for computing the relative contributions of uncertainty from GCMs and HMs. Many studies demonstrated that simple Analysis of Variance (ANOVA) approaches are perfectly suited to this case, and it has been recently widely applied to compute contribution from GCMs and HMs (see e.g. Giuntoli et al., 2015; Vetter et al., 2017, among many others), even by some of the authors of the present manuscript (Mishra et al., 2015). ANOVA approaches can critically take account of GCM/HMs interactions, which is presumably not the case of the method used, and of the different sizes of fixed effects. The set-up is here rather simple compared to more complex ones that consider unbalanced number of runs from each GCMs and/or multiple sources of uncertainty (see e.g. Addor et al., 2014; Vidal et al., 2016). I therefore strongly recommend using a simple two-way ANOVA approaches for the present study, or at least check current results against a simple two-way ANOVA approach. Indeed, I am unsure of how this sequential sampling approach relates to the more traditional ANOVA approach, and what their respective underlying hypotheses are. I would welcome some online discussion on this.

The rationale of not using ANOVA and the description of the sequential sampling procedure similar to that proposed by (?) was explained in (?). In short, standard parametric procedures, such as the Analysis of Variance (ANOVA), require assumptions of normality to estimate significance levels (the F and the t-student test require that the underlying variable is normally distributed). The low-flow statistics estimated in this study are non-normal and hence standard methods are not appropriate. For the estimation of the relative contributions of uncertainty from GCMs and HMs we use the range of the ensemble instead of the variance as suggested by Schewe et al. The confidence interval and the significance level of variability is estimated, in this case, with a non-parametric method (basically it is bootstrapping). This method is called sequential sampling in Samaniego et al. 2016. because it includes a "sampling with replacement" procedure to generate confidence intervals for the range statistic. Moreover, a non- parametric (bootstrapping) procedure is preferred here to reduce the

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effects of the biased variance estimation due to the small sample size.

20. P8L13-15: First, this should come much earlier in the manuscript. Second, this is not consistent with maps of streamflow changes that seemingly include results for catchments with a surface lower than 10000 km². This should be clarified. This is closely linked to specific comment #14.

First: Agreed. Second: explained in 1.2. The information that river grid cells with a contributing area >1000 km² are used in the results section will be included.

21. Figure 3 (and Fig. 5 and Fig. 6). See comment above. Plus, the figure indicated above each map is seemingly a continental average of the plotted value along the river network. First, this should be clarified. Second, this value is closely related to the choice of the minimal catchment surface area considered. Values would be very different if, as stated, only catchments with an area larger than 10000 km² would be considered. Please make all these statements and results consistent across the manuscript.

Agreed.

22. P10L3-4: This statement is somewhat inconsistent with the choice of the calendar year use for the calculation of Q90. Please clarify this in the manuscript.

Agreed. See #16

23. P10L7, "models": I presume this should be "simulations".

Yes, right.

24. P11L1, "new spatially explicit information". This is again contradictory with the 10000 km² statement. Cf. comments above.

See general comment 1.2

25. P11L16-17. This sentence is ambiguous. The increased spread along the 1:1 line (i.e. when smaller and larger values are considered) does indeed contribute to a higher coefficient of determination, which is not the case for the spread across (i.e. with higher residuals from) the 1:1 line. Please rephrase.

Agreed.

26. Figure 4: Several presumably regression lines are given on the graph. Please

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either define and comment them, or remove them. Also, please add lines delimiting the quadrants.

The regression lines are shown for positive and negative values. Modifications will be done as proposed.

27. Figure 4: The legend states that only catchments with a surface area higher than 10000 km² are considered. This is again not consistent with values provided by other figures.

This is true and was done with respect to the clearness of the figure. A comparison to surface areas higher than 1000 km² showed similar results.

28. P13L3-5: This is already written P10L35-P11L2. And this is commented in specific comment #24.

See #24

29. Title of Section 3.2: The difference between section 3.1 and section 3.2 are not understandable based on this title, and the reader may be unsettled at this point as I was. There should be something of a "between the levels of warming" somewhere. Please rephrase.

Agreed.

30. Figure 5. Cf. comment #21.

Agreed.

31. P15L15-16: The increase in winter low flows would not necessarily lead to a higher hydropower potential. It actually depends on the evolution of total precipitation. And the possible evolution of hydropower production would depend on the type of reservoir management, as well as management rules constrained by possible other water usages (sustaining summer low flows downstream, irrigation, recreation, etc.). Moreover, a decrease in low flows does not necessarily imply a decrease in overall water availability average over the year, and the water stress is conditional on the respective weight of water availability and water demand for a given time. So I would recommend adapting the statements according to the above comments.

Agreed.

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32. P15L17-18: I however completely agree with the need of regional adaptation options. Except that adaptation strategies should be put in place now, without waiting for the 3K level to be reached or not.

Agreed. We will include a sentence on this.

33. P16L6, "the result is independent of the sign of change": Well, this is a potentially serious issue. Indeed, how to interpret a situation where e.g. out of 15 projections, 5 give a significant upward change, 5 other no significant change, and the last 5 a significant downward change? I would recommend interpreting this situation with particularly no robust signal! So please make clearer in the manuscript all the different possible cases and the way to interpret them. An alternative for presenting robustness would be the one used in the IPCC AR5 WGI report, i.e. the percentage of projections agreeing on the sign of the change.

Fully agree with first statement. This is the reason why we suggest to use a combination of SNR and robustness. E.g. using the IPCC AR5 WGI report would give an information similar to SNR, but without a significance information.

34. P16L11-13: I totally agree with this sentence, but it comes here out of the blue. Please consider moving it to the introduction, discussion, or conclusion.

The sentence will be added in the conclusions.

35. Figure 6. The choice of colour breaks is here particularly unfortunate here. For the SNR, I would appreciate having a break in value 1, in order to see where the median change is higher than the uncertainty in projections. For the ratio of GCM to HM uncertainty contribution, this is all the more important to see where this crosses the 1 value. An alternative would be to use bivariate colour scales (Teuling et al., 2011) to jointly plot the evolution of both sources of uncertainty.

We understand that it is naturally to expect color breaks at 1 and we also used these at a previous version of this Figure. The purpose of this section is to highlight the substantial uncertainties associated with the results. For this reason, we decided to use a range of plus/minus 20% around 1. to mark regions where the contribution by GCMs and HMs (GCM/HM contr.) are of the same order of magnitude (please note

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that 0.8 and 1.25 are the inverse of each other for multiplication). It is misleading to distinguish a value slightly higher than 1. from one slightly lower (e.g., 1.04 from 0.98). Given the uncertainty in the analysed dataset, we only consider a higher contribution by either GCMs or HMs if it is at least 20% higher than the other. Similar arguments hold for the signal to noise ratio. We added a sentence to clarify these points (see p. 15 line 10). Using a bivariate color scheme following Teuling et al. 2011 is a possible alternative for the presentation of Figure 6. This color scheme would show the values in absolute terms rather than relative to each other. It would be possible to distinguish high and low values, but it would be harder to see which of the two sources of uncertainty is higher. We also think that showing the signal to noise ratio already allows to identify regions with high and low uncertainty and, additionally, providing absolute values is not required. Proposed paragraph to include: "It is worth noting that we have chosen the color scheme in Figure 6 in a way that regions where the SNR is within 20% around 1. have the same colors. These regions have a signal, which is of similar magnitude as the uncertainty. Different colors are used to mark regions where the signal is more than 20% higher or lower than the uncertainty."

36. P17L4-5: This exact sentence has already been written P8L3-4, and commented above (comment #19)

The sentence will be rephrased.

37. P17L8-P18L3: I am more or less OK with what is written here, but I do not understand why this would imply that the ratio of HM contribution to GCM contribution is higher at the 3K level. Please provide some explanations in the manuscript. Couldn't this be related to timing of threshold crossing in HM behavior that would differ from one HM to another, e.g. going from energy-limited to water-limited evaporation process?

Answered under #38

38. P18L4-20: This whole paragraph tends to support the above hypothesis. This should be related in the manuscript to recent uncertainty decomposition results obtained for a catchment located in the Southern Alps. It showed that the increasing spread of changes in future low flows by different HMs is linked to increasing spread

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in simulated evaporation and snow water equivalent (Vidal et al., 2016).

We agree with the reviewer that the explanation that we provided is only valid for the increase of the uncertainty of GCMs and that other factors such as the one mentioned by the reviewer influence the increase of HM uncertainty. We rephrased this paragraph to be more explicit about the different sources influencing the uncertainty contribution paragraph starting at p. 17, l. 7:

Total uncertainties in low flow projections is separated into GCM and HM contributions using the sequential sampling method proposed in Schewe et al. (2014). The results are shown in Fig. 6 (d-f) and spatially aggregated over the IPCC Europe regions in Tab. 4. The uncertainty rises with higher levels of warming for both sources of uncertainty because of two reasons. The GCM uncertainty increases because a 30 year period reaching a 3 K warming often has a strong temperature period within this 30 year period. Contrarily, GCM runs under RCP 2.6 often stabilise around 1.5 K global warming. This pathway dependency of warming influences the variability of the results with expectedly higher variability in the former case (James et al., 2017). The HM uncertainty increase with global warming because certain regions might cross thresholds. For example, parts of France might move from a energy-limited to a water-limited regime. Overall, the contribution of the GCMs to the uncertainty over Europe is about 21% higher under 1.5 K, 25% higher 10 under 2 K and only 10% higher under 3 K global warming in comparison to the HM contribution. This decrease of GCM/HM contribution can be mostly attributed to the Mediterranean and Atlantic region (in particular France). In these dry regions, the different representations of evaporation using temperature-based potential evapotranspiration used in mHM and PCR-GLOBWB lead to different responses than explicitly solving the full energy-balance of the land surface as in Noah-MP.

39. P19L28-30, “We conclude. . . support the adaptation process.” Well, this is actually only a wish. Nothing in the paper allows asserting that, even I personally hope this is the case. So please rephrase.

Agreed.

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3 Technical corrections

Technical corrections hereafter will be addressed according to the reviewers suggestions.

1. P1L5, “unprecedented”: it is a bit far-fetched, given that (1) GCM forcings are only disaggregated to this resolution without adding any downscaling information, and (2) results are seemingly partly given only for catchments >10000 km² (P8L13- 15).
2. P1L6: “combination”
3. P2L2: “independently”?
4. P2L22-24: I believe that the sentence is not grammatically correct.
5. P2L30: “2”? in reference (UNFCCC, 2015)
6. P2L34: “because of”
7. P3L3: please check missing or incorrect “the”
8. P3L8: “southern Europe”
9. P11L11: “extent”
10. P13L5: “political” -> “policy”. Also P15L23.
11. P15L22, “distinguished”: please rephrase.
12. P15L24, “ensemble members”: Please clarify what they are.
13. P20L1, “pronounced”: What is? Please rephrase.
14. P21L5-8: Wrong formatting, cf. IPCC report citation rules.
15. P23L34: line feed
16. P24L25-26: extra information to be removed

4 References

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