# Rebuttal Letter for the Reviews on 'Machine learning methods to assess the effects of a non-linear damage spectrum taking into account soil moisture on winter wheat yields in Germany’ 

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#### Abstract

We thank the reviewers for their detailed comments, which helped us to improve our manuscript substantially. In particular, the main comment from reviewer 1 was was very helpful because the corresponding analysis of why the clusters are identical to the state boundaries helped us identify an issue within the data provided to the clustering algorithms that was not valid. Unfortunately, this analysis is the basis of the statistical analysis and therefore has profound implications for all of the results presented in this manuscript.

We must apologize that we were not able to create a track change file due to the many changes in the tex-files. Here you will find the main changes in the tables and figures. The text also changes accordingly: As a result of the error, most of the changes can be found in Table 2 and Fig. 2, 3, 4, 5, A3, A4, and A5. In Table 2, the validation criteria associated with the cluster algorithm and size combination and the soil moisture depth used have been changed. Now the difference between the different soil moisture configurations is slightly smaller. Also, the prediction capacity shrank a little. Accordingly, a new cluster algorithm and size combination (PAM with cluster size 4) is used and shown in Fig. 2. There, the clustering for all other clusters can now also be found in Table 2. Fig. 3 now shows the scatter and density plots for four clusters accordingly. This also applies to Fig. 4, where the ALE plots for 4 clusters are now shown (only nine ALE plots per cluster instead of 12 before). This made the heat effect for the cluster in the northeast of Germany more relevant than before. However, many sensitivities are consistent with the results presented previously. Fig. 5 was created with the new model and cluster for PAM (4). However, the results there have hardly changed. In Fig. A3, the results for the cluster internal validation criteria have changed slightly. Fig. A4 now shows the relative importance for the non-cluster and the respective four clusters of PAM (4). Fig. A5 now shows the ALE plots for the hierarchical cluster algorithm with two clusters, i.e., those with the highest predictive capacity. We have now also added the ALE plots for the clusters shown in Table 2, which consider either the top 25 cm of the soil layer or both the top 25 cm and the entire soil column.


## Reviewer 1

## Summary

The authors present an assessment on the relationship of wheat yield anomalies to hydrometeorological features using cluster analysis and random forest together with accumulated local effects. They achieve good model performance and additionally present an in-depth investigation of the most important driver variables, which adds additional value to the article. They highlight that their results serve well for the regional identification of harmul seasonal hydro-meteorological conditions. Furthermore, they state the potential usage of their research e.g. for the identification of harmful features and their related thresholds and also carefully underline the limitations regarding out-of sample predictions, underestimations of extremes and missing inclusion of interaction effects in the model. I suggest that the manuscript should be accepted with moderate revision.

Dear Reviewer, thank you for your detailed summary and suggestions for accepting the manuscript with moderate revisions. We have responded to the major comments and the minor comments throughout the manuscript. Please find our responses below.

## Major comments

The fact that the clusters are identical with the borders of the federated states of Germany (or groups of them) without a single exception (as far as I can see) is not discussed and does not seem obvious to me. The authors should provide an explanation for this, e.g. stating which of the included variables in the cluster analysis are so strikingly different between the federated states that it leads to a perfect match of clusters and federated states (while any environmental / climatic differences do not matter in the clustering). It seems plausible that differences between eastern and western Germany due to the different political systems in place in the past can have a significant influence, but I wonder why this should hold true at the administrative level of federated states. The variables included in the cluster analysis are average yield and monthly averages and daily observations of the meteorological data for the entire year and SMI for both the upper layer and the entire soil column, right? I cannot see how these variables should primarily be superposed by the shapes of the federated states of Germany, even though I understand your argument that average yield is connected to farm size (which differs strongly between eastern and western Germany). We sincerely thank you for this comment, as the corresponding thorough review of the cluster approach helped us identify a problem within the data provided for the cluster algorithms. Unfortunately, the data contained a variable that included state information, which was an artifact from the previous processing of the data. This data is not scaled and thus dominates the clustering results. For this reason, the resulting areas correspond to the state boundaries. We sincerely apologize for this error. It unfortunately affects all subsequent results as well.

## Minor comments

[l. 11] " $R$-sqared" or preferably " $R$ ". Make this consistent throughout the paper.
Thank you for bringing attention to this unsightly habit. We use now " $\mathrm{R}^{2}$ " troughout the text.
[l. 16] "I assume you mean "crop yield variations".
65 Yes, we do.
[l. 79] Is the trend statistically significant?.
Yes, the trends in all three configurations (loess is defined as natural splines with $\mathbf{4}$ degrees of freedom) are significant. An ANOVA test shows, that the loess model with non-linear trend is significantly better than the linear model.
[Figure 1, 2, 5, A1] There are various grey counties, I assume mostly cities and/or counties without nonirrigated arable land
(Peichl et al. 2018). If all these counties are the ones with 0 years of data, this should be clearly visible and stated in Fig. A1.
Thank you, we adapted the Fig. A1 and the description accordingly.
[Figure 1 caption] Consider specifically naming the exceptional years 2003, 2014 and 2018 here.
Thanks, the years are now named.
[Table 1] It should say "max. $T>3^{\circ} C$ " for alternating frost according to Gömann et al. 2015
75 Indeed, that is how we defined it. Thank you for pointing out this error to us.
[Table 1] The presented variables in Table 1 seem plausible, but I wonder how excactly you came up with them? You sometimes depart from the months suggested by Gömann et al. 2015, e.g. they state the usage of Jan-Apr for Alternating Frost, while you also included May.

Thank you for drawing attention to this discrepancy. This research project also incorporated expert knowledge from interviews with farmers, which suggested the need to extend these time periods. Because the machine learning algorithm used here inherently performs feature selection, we therefore extended the time periods suggested by Gömann et al. (2015) rather than restricting them.
[Table 1] It would be beneficial to have all predictor variables for the random forest in one overview table, so consider including SMI here, too.

85 For a better overview, all variables are now included in Table 1.
[l. 100] "each grid point depends"
Thanks, we use cells instead of points to improve clarity here.
[l. l. 101-102] You compare your soil moisture index to simulations, but the soil moisture index used here is also simulated, so I do not understand this sentence fully. Could you elaborate please?
90 Soil moisture is presented here as an index because an index configuration supports the reduction of systematic errors of data that are simulated as well as spatially processed, such as in the present study (Auffhammer et al., 2013; Lobell, 2013)."
[l. 113, 350, 368] adjust citation brackets
Thank you for pointing this out.
[l. 113:] I understand the line of argumentation of the authors, that soil moisture is a slow-responding variable with a long memory, but relating this to autocorrelation per se might be a bit misleading here, because temperature is of course also auto-correlated. So consider writing something like "The autocorrelation / long-term temporal persistence of soil moisture is comparatively high in comparison to temperature".

Thank you, we rewrote the entire passage to clarify the points you made. Cumulative measures are widely used for meteorological variables as for instance growing degree days and killing degree days (Schlenker and Roberts, 2009) . [l.116] State here also specifically that on the other hand for SMI all months of the growing season were used (only SMIa is reduced to four months).

This is elaborated here and shown in Table 1 now.
[l. 119] My understanding is that masked grid cells are excluded, but you mean the opposite that you kept only non-irrigated agricultural land, right?
Grid cells that are not non-irrigated agricultural land are excluded and the remaining cells are used to calculate the respective county average.
[l. 145] Did you run the random forest model for each cluster size from 2 to 16 for each algorithm as you did for your standard internal validation?

Yes, we ran it for all possible combinations of cluster size (2-16) and clustering algorithms.
[l. 147-49] I wonder whether the statements on the data inlcuded in the cluster analysis should be rather mentioned adjacent to the data included in the random forest model in section 2 to have the complete overview at once.
We placed the data information here to avoid confusion with the data used in the Random Forests.
[Table 2] While SMIa stands for the entire soil column, usage of the term SMI is ambigous. You use it to refer to the uppermost 25 cm (e.g. l.111), but also as a general term for both depths (e.g. l. 148, Table 2). Consider using separate terms

We now use the abbreviation SMI only for the top 25 cm and when it is a general term we do not use the abbreviation SMI but write soil moisture index.
[Figure 2] You use dark grey for missing data and light grey for cluster 2. Consider taking another color for cluster 2 to make it better distinguishable.

The figure is now revised and the gray areas should now be easier to see.
[Figure 2 and A1:] coordinate degree sign as superscript
Thank you, we have not noticed this yet as it happened when converting the respective files.
[l. 189] You mention structural differences between western and eastern Germany. Consider mentionening briefly here the large differences in farm sizes due to different political systems in place, as an international audience might not be aware of this.

Thank you, we added this information backed by data from the Agrarstrukturerhebung in 2016.
[Figure 3] Indicate that the regression line is shown in bold black in the figure description. I assume the ellipses are point densities? Please specify this.
We added those information in Figure 3.
[l. 227] I assume this reduction is to be compared with an average month, right?: "for each month with an SMI value lower than 0.125 compared to an average month."
By definition, an SMI of 0.5 indicates the 50 percent quantile of the empirical distribution of water content for that county and thus serves as a benchmark for the average month (assuming that the empirical distribution is symmetric). The value of $\mathbf{0 . 1 2 5}$ is taken here as representative of a threshold at which yield falls rapidly relative to average yield expectations. We have modified this section to make the interpretation clearer.
[Figure 4] Is there a reason why there are many more red dots for the SMI compared to Heat and PS in a given subpanel?
Meteorological features are discrete values (counting days above or below a predefined threshold), while SMI is a continuous feature between 0 and 1.
[Figure 4] You state for Fig. A4 and in the text section describing Fig. A5 that the average of 50 repititions is taken. Is this also the case here?

Thank you, the 50 in Fig. A5 refers to the number of repetitions of the permutation to define the variable importance. Instead of using the phrase interval size, we now use grid size in the contest of ALE-plots to avoid further confusion. The plot is based on the feature importance assessment with 50 repetitions and then the ALE plots are shown for the most important features. These plots are based on a grid size of $\mathbf{1 0 0}$.
[Figure 4 and A4] You chose an interval size of 100, whereas for Figure A4 it is 50. Why did you choose different interval sizes?

The grid size in Figure $\mathbf{A 4}$ of 50 was an artifact of previous procedures. The ALE plots shown here are also based on a grid size of 100 .
[236] SMI year-round. "entire" could be confused with the entire soil column.
Here, "entire" referred to the entire range of the SMI in March. This has now been corrected.
[l. 239] "led to large losses"
Thank you, changed accordingly.
[l. 241] "but need to be"
Thank you, we changed it accordingly.
[l. 243] I assume "crop yield potentials" might be more precise.
Thank you, to include "yield" is indeed more accurate.
[l. 245] "water governs"
The structure of the entire sentence was odd so we changed it: "The observation that the absence of water governs crop production in eastern Germany is in alignment with recent studies."
[l. 246] delete "this"
See reply above.
[l. 266] "eleventh day"
We changed it accordingly.
[l. 268] Why especially East Germany? It is valid for non-cluster and cluster 1, too, is it not?

Thank you for bringing this ambiguity to our attention. We have changed the entire paragraph there: "Our approach, which explicitly controls for plant water supply through soil moisture, generally shows (for the no cluster approach as well as cluster 1 and cluster 2) more negative effects in terms of water deficit compared to heat. Especially for eastern Germany (cluster 2), the water deficit in the upper 25 cm of the soil plays a prominent role in late summer and spring."
[l. 274] delete "years" in front of "in-sample"
We have deleted years.
[l. 274] It did not directly become clear to me that 2019 is an out-of-sample year at the first read. I suggest explicitly naming it an out-of-sample year in this context (as you also do below).

We added this information to clarify in the beginning of this paragraph that 2019 was not used in the training set.
[l.286] "in some of the easternmost"
Thank you, we changed it accordingly.
[l. 288] "the highest negative"
Thank you, we adopted this.
[l. 314] You use the terms East / eastern Germany and northeastern Germany. If all these mean the same region, consider using only one term or are you specifically referring only to cluster 7 with northeastern Germany?

Thank you, we changed the term here to eastern Germany. For the rest of the text we use eastern Germany when the spatial scope is the focus, and East Germany when the political system and the history of the region is of particular interest.
[l. 324] GCM: Abbreviations should be written out at first occurrence.
Thank you, we introduced global climate models.
[l. 355] "a small number" "a rather large number"
Changed accordingly, thank you.
[l. 367] Do you mean "mean absolute error"?
Yes, indeed, the mean absolute error is meant.
[l. 376/378] Do you refer to SMI11 and SMI12 as "lagged"? Simply because they belong to another calendar year does not make them "lagged" variables in my opinion, so this term might be inappropriate here.

Thank you, to avoid any confusion when it comes to those terms we instead refer to the previous years.
[l. 377] Is SMI11 or Heat8 meant here?
SMI11 of course.
[Figure A4 caption] "50 repetitions"
Thank you, we adopted this.
[Figure A5 caption] The caption should specify that the range of each variable stems from 50 (or 100?) repititions.
Thank you, we included this information now.

## Reviewer 2

General comments

This paper presents the modelling of non-linear effects of meteorological divers and soilmoisture on winter wheat yield variability. A random forest procedure models the nonlinear relationships. The model is applied on subregions of Germany, obtained with a clustering procedure. A comparison with the model trained over the whole country emphasizes the relevance of the clustering. The authors highlight the importance of soil moisture as a relevant explicative variable, more relevant than heat. The manuscript is well written. The description of the results is clear and supported by existing literature. This paper deserves publication after minor modifications/addition of complementary information.
Dear Reviewer, thank you for your detailed summary and suggestions for accepting the manuscript with minor modifications/addition of complementary information. We have responded to the specific comments and technical corrections throughout the manuscript. Please find our responses below.

## Specific comments

[Table 1] The description of variables is generally self-explanatory, except maybe "alternative frost". Does it refer to the number of consecutive pairs of days with $\min T<-3$ and then $\min T>3$ ?

Thank you for drawing attention to this need for clarification. First, the minT $\mathbf{>} \mathbf{3}$ should be declared as maxT $>\mathbf{3}$. Also, it takes into account days when both conditions apply, i.e. a day with, for example, minimum temperatures below $\mathbf{- 3}$ degrees at night and then during the day with degrees higher than $\mathbf{- 3}$ maximum temperature.
Is there a correlation between SMI and SMIa, and can this have an impact on the quality of the RF model? Same question for correlation between $\operatorname{SMI}(a)$ and indicators such as Heat, Heavy rain, precipitation scarcity.

There is a correlation between SMI and SMIa as well as with all meteorological variables. However, it is generally understood that this does not affect the predictive capacity of Random Forests - and this is what we assume is meant by model quality here. Nevertheless, it does affect the interpretability of the model. In our case, this is in particular the case for feature importance (if the are multiple features that contain the same information this as an impact of the order of partitioning the data). Since we are using accumulated local effects, purged of any correlation, this should be less affected by correlation issues such as multicollinearity.
[l.118/119] SMI is masked for non-irrigated agricultural land, but are these areas also discarded in yield data?
As the yield information are only available on administrative district level this masking was not possible. When it comes to the factor irrigation, we consider this neglectable as only about 5 percent of the agricultural area is possibly irrigated with an focus on crops like potatoes (https://www.destatis.de/DE/Themen/Branchen-Unternehmen/Landwirtschaft-Forstwirtschaft Fischerei/Produktionsmethoden/Tabellen/bewaesserungsmoeglichkeiten.html)
[l.138] How can one interpret quickly subregions within Germany obtained with clustering? Are they areas where yield is of the same order of magnitude and also monthly and daily meteorological are also similar?

A more detailed description can be found in our response to the main comment of reviewer 1. Unfortunately, the cluster algorithm is provided with invalid data regarding the federal states that impact the cluster formation.
[Figure 2] Please specify in the caption or the legend that the numbers in rectangles are referring to the Rsquare obtain from the RF procedure(?).

The entire figure now has been revised. Now, we state in the caption that the number indicate the respective test $R$ squared.
[Figure 2a] Is it by chance that except cluster number 8, all clusters are simply connected and almost convex? What could explain this very smooth partition?

Please see reply above and to reviewer number one.
[l.186] Would it a better option to use PAM(3) (with only SMIa) than PAM(2) (with both SMI)? (To get rid of potential correlation problems between SMI and SMIa).

We apologize for the confusion caused by not further clarifying that we are relying only on the top $\mathbf{2 5} \mathbf{c m}$ in this setting. To fix this problem, we have added this sentence to the paragraph before it: "Since the data for the entire soil column do not appear to provide any additional information for the model, we rely only on the top $\mathbf{2 5} \mathbf{c m}$ for further analysis." [l.202] What would be a solution to avoid this overfitting of the model?
In this case, we compare the ALE plots of RFs fitted with either the PAM8 or the PAM2 setting. Thus, the former is clustered into eight subregions. This means that each RF model is trained on a smaller sample size such as in the PAM2 context or when no cluster is used. Consistently, this is also the model and corresponding sample in each case on which the ALE plots for that subregion are based. Because of the less smooth functional relationships shown for this configuration, we assume that the models are overfitted to this small sample size. One possible solution would be to fit to larger samples. This is done here by relying instead on PAM2, i.e., the setting that considers only two clusters and therefore has a higher sample size in each cluster.
[l.202] "The effects shown here are additive as they are cleared off the correlation to other features". I don't fully understand this sentence. Could you be more specific?

When features interact with other features in a prediction model, the prediction cannot be expressed as the sum of the feature effects, because the effect of one feature depends on the value of the other feature. Because of this compounding, the features are correlated to each other. In linear models this is commonly expressed by multiplicative expressions. As the ALE plots are purged of this correlation, we can rely on the sum of this features to derive the overall effect. This is similar to a linear model with only additive terms. We added some clarifications: "The feature effects shown here can be interpreted as additive because they are purged of correlation to other features. For example, the combined effect of soil moisture in June and July is the sum of SMI6 and SMI7."
[Figure 4] Do the black and white bars in x-axis represent the distribution of the explicative variable?
Yes, the black bars show the distribution of the respective features. White bars a results of discrete features.
[Figure 4., caption] It is not clear to me what the interval size of 100 refers to.
ALE plots predict the effect of an explanatory variable across their realisations, taking into account only a subset of the
sample with observed values adjacent to the respective realisation. The size of this subset of the sample is defined by the grid size. The larger the grid size, the smaller the subset and the less smooth the visualization of the average marginal effects.
[l240] Would it be possible to add simple interactions in the model? (multiplication of 2 variables?)
In general, it would be possible to add these interactions by simply converting the features themselves into interactions, e.g. by multiplication of two variables as suggested. However, due to the strongly nonlinear structure caused by the underlying recursive partitioning, we consider such an approach not necessary for (large enough) random forests. [l265-266] "In both clusters, heat in August, a period generally associated with ripening, has positive effects for each additional day and from day 11 onward negative effects" According to fig4, it looks like only for cluster 1, Heat8 has a negative effect from day 11 onward, not for cluster 2.
Thank you, we now distinguish between cluster 1 and cluster 2: 'In both clusters, heat in August, a period generally associated with ripening, has positive effects for each additional day and negative effects after the eleventh (cluster 1) respective sixth (cluster 2) day."
[l.291-295] Can the difficulties of out-of-sample prediction be interpreted as overfitting? Could it be Improved with longer time series (to have a larger number of configurations)?
The validation criterion the random forests are based on are out-of-bag error estimates. With tree sizes large enough the estimates converge to those found for leave-one-out cross-validation. This validation technique is more prone to overfitting compared to other methods. Potential solutions are the use of 1) an extended time-series, 2) the use of a different cross-validation technique with higher focus the variabilty of the prediction but less the bias in the training data, 3) the use of machine learning techniques that might be better suitable to extrapolated out-of-the sample compared to random forests. We clarified this in more detail in this paragraph and the conclusions.

## Technical corrections

## [l.113] Citation in brackets

Thank you, the citations are now corrected.
[l.154] Missing bracket
Thank you, we included the missing bracket.
[l.202] Extra "the" in "The effects shown here are additive as the they are cleared"
Thank you, we deleted the extra "the"
[l.377] Is "(Heat8)" supposed to be in that sentence?
No, it is supposed to be SMI11. Thanks for bringing this to our attention.
[Caption figure A4] "50 repetitions"
Thank you, it is of course 50 repetitions.

## References

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