

Interactive comment on “What do we do with model simulation crashes? Recommendations for global sensitivity analysis of earth and environmental systems models” by Razi Sheikholeslami et al.

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This document contains copies of all comments of the Reviewer 1 and our planned efforts to address them in the revised manuscript.

Reviewer 1: With great interest I have read and reviewed the manuscript “What do we do with model simulation crashes? Recommendations for global sensitivity analysis of earth and environmental systems models” by Sheikholeslami et al. In general, the paper presents a novel and interesting approach to deal with the issue of model

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crashes when applying global sensitivity analysis. Although this new idea looks promising, additional investigations and explanations are necessary before this paper can be published. In the next sections general, major and minor comments and suggestions are provided that should allow the authors to improve their manuscript.

Response: We are very thankful to the reviewer for the time and effort spent on reviewing our paper. The comments and suggestions were constructive and helped us improve the quality of our manuscript. We will address all the comments in the revised manuscript as described in this rebuttal document.

Reviewer 1: General. To improve the validity of this novel idea and allow it to be applied in a more general context, more investigation is required by:

- Applying this for different SA techniques (e.g. a variance-based technique, as one of the proposed approaches might influence the variance of the output) (in particular on the HBV example with different ratios of the number of crashes).

Response: Thanks for this important comment. As suggested by the reviewer, we have tested the proposed crash handling strategy using a variance based GSA technique and compared the results for the first case study. We will add these results into the revised manuscript. However, note that our proposed strategy is GSA method-free and does not depend on the utilized GSA algorithm.

- Applying the k-NN technique instead or next to the simple NN, as the former seems to be more powerful.

Response: We agree that the k-NN technique is an effective regression approach compared to the single NN. Considering our main goal in this study, i.e., finding simple yet effective strategies for handling simulation failures, we have adopted the single NN technique because it is a simple, persimmons method. Of course, more complex metamodeling options are available, including k-NN. The objective was not to provide a comprehensive comparison though. We have added the following paragraph in Section

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2.2.2 to address this comment:

"The single NN is a parsimonious, very simple to understand, and equally easy to implement. To fill-in the crashed simulations, single-NN algorithm reads through whole dataset to find the nearest neighbor. On the other hand, the k-NN technique is more complex and requires a careful selection of the kernel functions to assign weights based on the degree of similarity between nearest neighbors. The choice of kernel functions and variable k is subjective. Particularly, k is a hyperparameter that user must pick to get the best possible fit for the dataset. However, the optimal k will always vary depending on the dataset. It is noteworthy that some authors have asserted that covariances among Y variables are preserved in the NN-based techniques when using small k-values (Hudak et al., 2008; LeMay & Temesgen, 2005; McRoberts et al., 2002; Tomppo et al., 2002). McRoberts (2009) showed that the variance and covariance of the Y variables tend to be preserved for $k = 1$ but not for $k > 1$ (McRoberts, 2009)."

- Applying a convergence analysis on the SA results. It appears to me that the proposed approach only slows down the convergence of the results (so an evolution of the SA statistics for both the simulations without crashes and the simulations with suggested crashes should be performed. Possible approaches can be found in (Sarrazin et al., 2016) or (Nossent et al., 2011)).

Response: We think there is no point in this comparison, as when a model crashes the classic convergence analysis does not make sense because the GSA algorithm cannot be finished. In other words, since crash strategies are applied after all runs are completed and without any need to repeat any experiment, crash strategies should not matter for convergence. To emphasize on the importance of the convergence analysis for the GSA, the following paragraph has been added into the discussion section (Section 5.2):

"It is important to note that the sample size in GSA studies should not only be determined based on the available computational budget but also considerations of GSA

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stability and convergence. Therefore, it is of vital importance to monitor and evaluate the convergence rate of the GSA algorithms. Strategies introduced by Nossent et al. (2011), Sarrazin et al., 2016, and more recently by Sheikholeslami et al. (2019) enable users to diagnose the convergence behavior of the GSA algorithms."

- Applying different sampling techniques (as the density of the samples might have an influence on the results) (e.g. on p13, L33 one could argue that this statement should be supported by applying different strategies next to "STAR").

Response: Thanks for pointing out this comment. As mentioned in Section 5.2, regardless of the chosen method for handling simulation crash problem in GSA, it is advisable to spend some time up front to find an optimal sample set before submitting it for evaluation to the computationally expensive models. Therefore, we used an advanced sampling strategy called Progressive Latin Hypercube Sampling (PLHS) to ensure sufficient coverage of the parameter space (please see Section 2.3). In other words, the STAR-VARS algorithm employs the PLHS strategy to locate star centers in the first phase of sampling.

- Adding information on the computation time of the different step.

Response: To address this comment, the following sentences have been added to the revised manuscript:

"The entire set of 100,000 function evaluations of the MESH model will take more than 6 months if a single CPU core is used. However, we used the University of Saskatchewan's high-performance computing system to run the GSA experiment in parallel on 160 cores. Therefore, completing all model runs required approximately 32 hours."

"For the MESH model, using an Intel® Core™ i7 CPU 4790 3.6GHz desktop PC, the RBF technique took only 65 seconds to substitute 3,084 crashed runs, while the single NN technique required about 197 seconds to complete the task."

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Reviewer 1: Major comments:

* p2, L27: In most cases, the samples for GSA are independent. This is important in the interpretation of your proposed strategy, so you should clearly mention this in the text.

Response: Thanks to Reviewer 1 for this important comment. As we mentioned in the introduction section, in case of model crashes, re-running the entire experiment is inevitable when using a GSA technique that utilizes a sampling strategy with a particular structure. To address this comment, we have revised Section 1.2 as follows:

"Ignoring the crashed runs in GSA is only relevant when using purely random (and independent) samples (i.e., Monte Carlo method). In such cases, if the model crashes at a given parameter set, one can simply exclude that parameter set or repeat randomly generating a parameter set (at the expense of increased computational cost) that results in a successful simulation."

"Reducing the number of model runs and finding optimum locations for sample points in the parameter space are the main drivers for implementing improved sampling techniques in GSA. Typically, these sampling techniques are not only random but also follow specific spatial arrangements. However, when applying a sampling-based technique that uses an ad-hoc sampling strategy with particular spatial structure (e.g., the variance-based GSA proposed by Saltelli et al. (2010) or STAR-VARS of Razavi and Gupta (2016b)), we cannot ignore crashed simulations. In this case, excluding sample points associated with simulation crashes will distort the structure of the sample set, causing inaccurate estimation of sensitivity indices. As a result, the user may have to re-do a part or the entire experiment depending on the GSA implementation, by generating a new sample set (or a succession of sample sets), leading to a waste of previous model runs."

* p4, L22: In many cases of GSA, it is not necessary to re-run the entire experiment, but just a limited number of runs. This is important to put this into perspective.

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Response: Agreed. Please see our response to the previous comment.

* p10, L11: What about parameter “CO”? It is influential, but you don’t talk about that one.

Response: Thanks for pointing this out. To address this comment, we have edited the text as follows:

"From the results, we see that using the RBF technique the sensitivity indices of the most important parameters (FRAC, FC, C0) (Fig. 4(a)) and less important parameters (LP, ETF, beta, K2) (Fig. 5) were estimated with high degree of accuracy and robustness."

* p12, L18: I have the impression that you went into detail too much on these causes of crashes of this model, whereas your main focus should be on the SA. The part starting on p13, L10 can be maintained as it is an interesting addition to this topic.

Response: As suggested by the reviewer, the length of this section has been reduced in the revised manuscript.

* p13, L26: Is this valid for both single NN and k-NN? Specify this.

Response: Yes, the poorly sampled parameter space can also influence the performance of the k-NN technique. In regions of the parameter space where the sample points are sparsely distributed, distances to nearest neighbours can be high, leading to choosing physically incompatible neighbours. To clarify this point, we have modified the text as follows:

"For example, in the NN techniques (both single and k-NN) one major concern is that the sparseness of sample points may affect the quality of the results."

* p25, fig 7: Could you provide some additional figures of this type in annex? Although this is arbitrarily chosen, this would support the results.

Response: We agree with the reviewer that reporting these figures might be of some

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value but adding these results would not change the discussions and conclusions already presented, and so we prefer not to add more figures to our already long paper having many figures. To further investigate the performance of the single NN and RBF techniques, we have added a new figure (the bottom panel of the Fig.7) into the revised manuscript.

Reviewer 1: Minor comments:

Response: Thank you very much for editing and proofreading our manuscript.

* p1, L28: Should it be “Dynamical Earth Systems Models” or “Dynamical Earth System Models”?

Response: As suggested, we used “Dynamical Earth System Models”.

* p2, L29: Remove either “that” or “how”

Response: Fixed.

* p3, L21: Replace “is” by “are”

Response: Typo fixed.

* p3, L24: Add “the” before “parameter space”

Response: Fixed.

* p4, L24: Replace “the” by “a”

Response: Fixed.

* p5, L21: It would be either “a computationally simple method” or “computationally simplest method”

Response: Corrected.

* p6, L18: Add “a” before “response”

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Response: Fixed.

* p6, L20: Add a comma after “In the literature”

Response: Fixed.

* p6, L22: Add “the” before “RBF”

Response: Fixed.

* p8, L7: Add “a” before “highly”

Response: Fixed.

* p8, L8: Add “a” before “minimum”

Response: Fixed.

* p8, L15: Add “a” or “the” before “maximum”

Response: Fixed.

* p8, L16: Add “the” before “output”

Response: Fixed.

* p8, L16: Add “a” before “minimum”

Response: Fixed.

* p8, L24 (and others): All superscript numbers seem to be written as normal numbers

Response: Typo fixed.

* p8, L26: “of which” should go before “10”

Response: Corrected as suggested.

* p8, L29: The last sentence seems to have an odd structure

Response: It has been modified.

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* p9, L20: Add “the” before “STAR-VARS”

Response: Fixed.

* p9, L25: Add “the” before “GSA”

Response: Fixed.

* p9, L30: I would suggest to move “when there are no crashes” between the brackets on the previous line (“after 9100 function evaluations”).

Response: Corrected as suggested.

* p10, L7: Add “the” before “parameter space”

Response: Fixed.

* p10, L7: Remove the “s” from “ratios”

Response: Typo fixed.

* p10, L10: Reformulate this sentence

Response: We have modified p10, L10 and now it reads:

"Moreover, Fig. 5 shows that when crashes were substituted using the RBF technique, the STAR-VARS algorithm estimated the sensitivity indices of the most important parameters (FRAC, FC, C0) (Fig. 4(a)) and less important parameters (LP, ETF, beta, K2) (Fig. 5) with higher degrees of accuracy and robustness."

* p11, L1: Add “the” before “Four” and before “water”

Response: Fixed.

* p11, L10: Replace “with” by “between these”

Response: Corrected.

* p11, L12: Add “a” before “vegetation”

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Response: Fixed.

* p11, L13: Add “the” before “soil”

Response: Fixed.

* p11, L21: Reformulate “As shown”

Response: It has been edited as follows:

"The STAR-VARS algorithm identified these parameters as weakly influential (very low IVARS-50 values) using the proposed crash handling techniques. However, the associated sensitivity indices obtained by the RBF imputation method are two orders of magnitude larger for the parameters in the left panel (Fig.11 (a, c)) and about four orders of magnitude larger for the parameters in the right panel (Fig. 11 (b, d)) compared to those obtained by the single NN and median substitution methods."

* p11, L23: Add an “s” to “order”

Response: Fixed.

* p11, L29: Remove the “7”

Response: Fixed.

* p13, L30: Replace “depends” by “depending”

Response: Typo fixed.

* p14, L1: Which feature? Reformulate this sentence

Response: Corrected.

* p14, L6: Replace “are” by “should be”

Response: Corrected.

* p14, L7: Add an “s” to “problem”

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Response: Typo fixed.

* p14, L25: Remove the “s” from “involves”

Response: Typo fixed.

* p14, L31: “The efficiency of our proposed simulation based strategies was shown: :
:.”

Response: Fixed.

* p14, L30: This is a very long, complex sentence.

Response: This sentence has been reformulated in the revised manuscript and now it reads:

"The efficiency of our proposed substitution-based strategy was shown to be remarkable compared to other crash handling strategies (ignorance-based and non-substitution procedures). This is of prominent importance particularly when dealing with GSA of the computationally expensive models mainly because the proposed strategy does not need repeating the entire experiment."

* p15, L11: “causing” instead of “casing”

Response: Typo fixed.

* p15, L18: “understanding” instead of “understand”

Response: Typo fixed.

* p23 caption: Remove “C0” from the list of “moderately influential parameters”

Response: Corrected.

Interactive comment on Geosci. Model Dev. Discuss., <https://doi.org/10.5194/gmd-2019-17>, 2019.

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