

REVIEW of the revised paper

Coupled machine learning and the limits of acceptability approach applied in parameter identification for a distributed hydrological model

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The revised version of the manuscript is a significant improvement over the initial version though I do not agree with some of the responses (see below). The manuscript can be published after minor revision.

I am not fully convinced with the response given by the authors about computational time saved by the emulators. It is not that critical to save computation time for offline simulations. Because calibration or training is generally done one time unless it has to be updated frequently due to significant change in input data distribution. The critical is to save computational time for real time application as mentioned in my comments on earlier version of this manuscript. The proposed method does not provide any benefit over the existing method particularly for real time application. This should be acknowledged at least in the discussion.

Page 4, Line 25: Replace “*The percentage of observations where model predictions fall within the limits*” with “The percentage of the model predictions that falls within the observation error limit”

Page 5, L4: “... *chosen certainty level (e.g. 5-95 %) based on previous experience or literature values.*” Provide references.

Page 3, L 15: prediction error: is this Observation-Simulation or vice versa. It is important to define as error is not absolute (according to response). The response given on page 9 (*Here, the notation e is not absolute and thus the expression $\mu_Q=0, e \leq L_e$ is correct, since a model producing a negative error value of less than the lower observational error bound (which is also a negative value) has 0 degree of membership*) does not make sense. Let us assume observation Q_{obs} is 100, then according 25% observation error, L_e is 75 and $U_e = 125$. Let corresponding simulation Q_{sim} be 70. According to equation 2, Since $Q_{sim} < L_e$, $S(Q_{sim})=0$, this is fine. Now if authors use same notations of L_e and U_e in equation 2 and Figure 1 and $e = Q_{sim}-Q_{obs}$ then problem arises for calculating membership of prediction error (See below)

- Case 1: simulation below L_e , e.g. $Q_{sim} = 70$, so $e = -30$ which is less than L_e , so membership = 0
- Case 2, simulation above L_e but below m , e.g. $Q_{sim} = 80$, $e = -20$, which is also less than L_e , so membership = 0
- Case 3, simulation below U_e , but greater than m , e.g. $Q_{sim} = 110$, $e = 10$, membership = $(125-10)/(125-100)$

- Case 4, simulation above U_e , e.g. $Q_{sim} = 130$, $e = 30$, membership not 0 because e is not greater than U_e

So notations L_e , U_e used in Figure 1 are not same as used in equation 2. In equation 2, notations should be something like this $L = Q_{obs} - 0.25 * Q_{obs}$, $U = Q_{obs} + 0.25 * Q_{obs}$. Then in Figure 1, it should be $L_e = L - Q_{obs}$ and $U_e = U - Q_{obs}$ which will satisfy membership function given in Figure 1. I strongly suggest to use notations of figure of earlier comments in the original version of the manuscript which is also consistent with equation 2.

Table 1: Provide size of S4 on Table 1.