

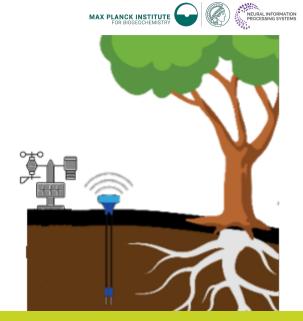
# **Ecosystem Insights through Extreme Values: A Fresh Look at Meteorological Drivers**

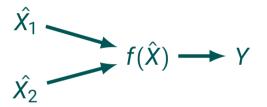
Jena, November 3, 2025

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# **Research Gap**

- The main purpose of data-driven models is predicting some response to specific inputs.
- For example, the amount of CO<sub>2</sub> uptake given meteorological forcings, or the amount of evaporation in a situation.
- However, in many disciplines researchers also build data- driven models in order to identify the correct mechanism.

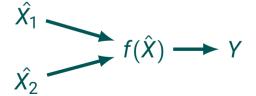




When we build a data-driven model, we want to identify a mechanism f(X) that determines some outcome Y from some drivers X<sub>1</sub> and X<sub>2</sub>.

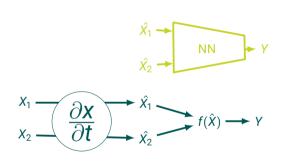






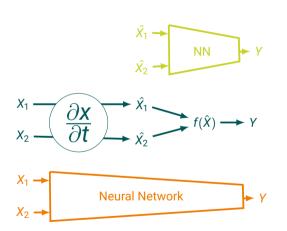
We can understand this mechanism by training a neural network.





- However, in real applications, we do not measure the inputs directly when they enter the mechanism.
- Instead, the variables interact between the point we measure and the point of entering the mechanism.



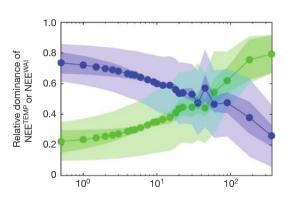


- Therefore, when building a data-driven model, we model the combination of the dynamic system and the target mechanism.
- Resulting in a biased prediction

## Where is this Relevant?



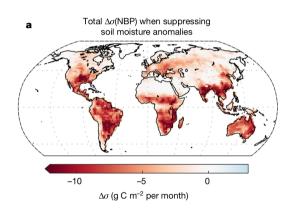
(Jung et al., 2017) find that plant productivity is locally mostly influenced by water availability but globally by temperature.



#### Where is this Relevant?



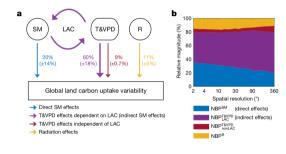
- (Jung et al., 2017) find that plant productivity is locally mostly influenced by water availability but globally by temperature.
- (Seneviratne et al., 2013) find that global models lose more than 80% of the variability in plant productivity when supressing the variability in soil moisture.



#### Where is this Relevant?



- (Jung et al., 2017) find that plant productivity is locally mostly influenced by water availability but globally by temperature.
- (Seneviratne et al., 2013) find that global models lose more than 80% of the variability in plant productivity when supressing the variability in soil moisture.
- (Humphrey et al., 2021) claim that this difference is due to the land-atmosphere feedback.



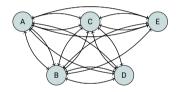
## **We Propose**



We propose to train only on extreme values.

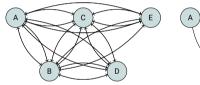
# Why?

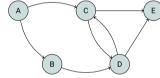




# Why?

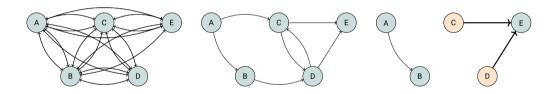






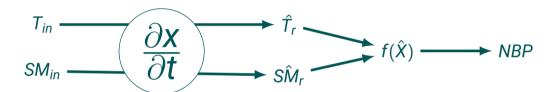
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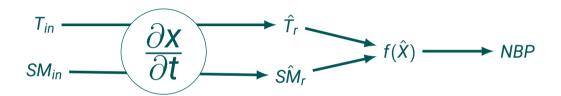
## Does it work?





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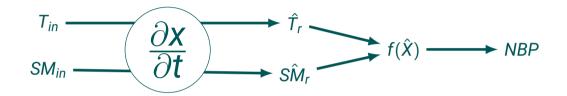




$$NBP = \alpha T_r + \beta SM_r + \gamma + \epsilon$$

## Does it work?





$$T_r = T_{in} - rac{C_T}{1 + e^{-(SM_{in}T_{in} - C_0)}}$$
  $SM_r = SM_{in} - rac{C_{SM}}{1 + e^{-(SM_{in}T_{in} - C_0)}}$ 

$$\textit{NBP} = \alpha \textit{T}_{\textit{r}} + \beta \textit{SM}_{\textit{r}} + \gamma + \epsilon$$

## **Theoretical Result**



If we can make reasonable assumptions, a linear regression on  $T_{in}$  and  $SM_{in}$  does not return the correct values  $\alpha$  and  $\beta$  but

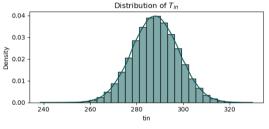
$$lpha^* = rac{lpha}{1 - C_T SM_{in}}$$
 and  $eta^* = rac{eta}{1 - C_{SM} T_{in}}.$ 

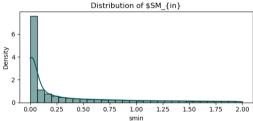
However, if we train only on extreme values, we find

$$\alpha^* = \alpha$$
 and  $\beta^* = \beta$ .

# **Empirical Results**

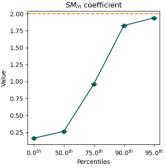


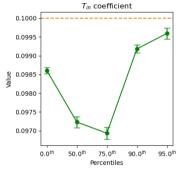


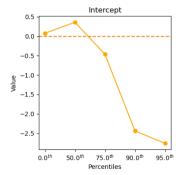


# **Empirical Results**









## **Conclusion and Future Work**



➤ Training on extreme values recovers the true mechanism while training on all values does not.

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- ► Training on extreme values recovers the true mechanism while training on all values does not.
- ► For the example we showed this theoretically as well as empirically.
- ► The next step is to test this with more realistic models and, afterward, on real data.