Closing the gap between statistical and scientific workflows for improved forecasts in ecology

Victor Van der Meersch, J. Regetz, T. J. Davies* & EM Wolkovich

March 27, 2025

1. Intro

- Ecology super challenged to predict stuff for decision making (kind of a whole new world of relevance?)
 - \rightarrow Example of stuff to predict: populations, policy-relevant questions...
- General way to do this so far... (bifurcated?)
 - long term data (GBIF, Biotime...) for estimation of trends
 - PBM, SDMs for forecast
- Gap, problem: None of this is going well
 - debates over trends, not only on the significance but also the direction!
 - predictive modeling trapped in overly complex models, hard to get scientific insights?
 (same as GCMs?)
- Here we introduce a universal workflow to address this (say more)

2. Overview

- General scientific method we all learn stresses: RQ \rightarrow study design \rightarrow collect data \rightarrow build model \rightarrow answer
- Divergences from this are common (Figure 1)
 - bad science (data lead to question)
 - important question for which we cannot get data quickly, we have to use existing data: e.g. global datasets for trends, many various small datasets for process-based (experimental...)
 - complexity makes the ideal workflow hard/impossible
- the workflow we present here works across these realities by
 - stressing the need to think about model before study design
 - advancing data simulation

^{*} Says he is happy to help and give friendly review, but not sure he will reach level of co-author.

- More details on ideal workflow: walk through the different steps (**Figure 2**), and emphasizes: not rigid structure, integration of everything, iterate the process of science, explicit effort to recognize the uncertainties,
 - spend more time in critical quadrat, post-model pre-data
 - feedbacks
 - uncertainty
- 3. How to address current issues (two case studies, **Figure 3**)

(a) Trends!

- Outline current problem: different answers from different analysis (and slightly different datasets?)
 - \rightarrow It is a problem because we can't make decisions on +/- trends debate... (and it degrades trust in science?)
- two big missing parts are data simulation steps:
 - retrodictive check: would highlight missing pieces (speed up current process, because without this step each slightly different model is a paper... whereas each big iteration of the workflow—including multiple feedbacks—should be a paper)
 - simulated data: you would know that you have low data so oner =; the debate is actually not jus models but really limited data

(b) Forecasts!

- Outline current problem:
 - black box: intrication between model build and data fitting (calibration), everything is mixed
 - complexity trap [mention uncertainty], counterproductive, not always better performance
 - developping a model has become the goal, whereas it should be a way to answer a research question!
- We need the workflow to open the black box! Simulating data would allow to add a necessary step between model building and data fitting, which would highlight strong degeneracies
- a clear framework to support (or not) additional complexity and new parameters
- Workflow would also force you to clearly express a research question, define a limited context in which the model should apply

(c) Step back

- we need more data, and better question (relate this to both previous study cases)
- where can we best reduce uncertainties through new scientific insights?
- machine learning! If we change nothing, what's the point of not doing ML? ML
 process-based without question, and ML > trends without mechanisms! Where theory is lacking, or where we are far from mechanistic understanding, you might as well do ML!

- 4. Wrap up: how to make it happen?
 - usual issues: publication pressure, low standards (especially for models)
 "we allow far more hand-waving in the presentation of modeling results than we do for
 experimental data" (J. Aber, 1997)
 - growing concerns, leading to increase in reproducibility and data sharing practices [mention uncertainty]
 - need a little more here...
 - better training! on BOTH estimation AND prediction
 - estimation: being aware of what is a parameter, and mention uncertainty propagation
 - prediction: should be a natural outcome, not a finality
 - ML (short-time forecast), benchmarking models are probably useful but should not be the core of our scientific practice, not the spirit!
 - \rightarrow moving ecology in the right direction!