Uncertainty: The Good, the Bad and the Beneficial

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Doubt is not a pleasant condition, but certainty is absurd. - Voltaire



Main Points

- Determining important sources of uncertainty for your project.
- Understanding uncertainty through ensemble modeling.
- Concept of relevant and useful uncertainty.

Taming the Uncertainty Monster

(Van der Sluijs, 2005; Curry and Webster, 2011)

- Monster Hiding
- Monster Exorcism
- Monster Simplification
- Monster Detection
- Monster Integration



What uncertainty monster do you deal with?



Sketch your uncertainty concept map

Soil moisture? Monsconrains Snowpack? Snowmelt? Streamflow extremes over the next 10 years? Forced Climate Climate ? Model error? Change? Variability? Model error?

Future Weather Extremes are Unknown

We don't know what the future extreme weather will be, but we know the range of possibilities and their likelihoods.



Future Weather Extremes are Unknown and Uncertain

If future extreme weather is a roll of the die:



Then our projections of future extreme weather are rolls of what we think that die looks like:



Many Uncertainty Sources







Model Uncertainty: Physics





Can your model generate the extreme phenomena?



Model Uncertainty: Sample Size

Tropical cyclogenesis locations from 1950 to 2003





Determine your sufficient sample size

Hall and Jewson (2007)

Boundary Forcing Uncertainty



Perhaps choose the strongest forcing to get a signal



Initial Condition Uncertainty



Need a large domain to sample initial condition uncertainty

The Dominant Uncertainty Varies with Scale



On regional scales, sample IC and model uncertainty



Multi-model approach samples greater uncertainty.

depends 6 D **Jncertainty** Φ



Can you use temperature instead?

Laboratory Task 1

What are the relative contributions to uncertainty in seasonal tropical cyclone (TC) numbers and intensity from internal variability and model physics?

TC track data:

- Single year (1998) 16-member initial condition ensemble.
- Single year (1998) 14-member model physics ensemble.



Done, J. M., C. L. Bruyère, M. Ge, and A. Jaye (2014), Internal variability of North Atlantic tropical cyclones, J. Geophys. Res. Atmos., 119, 6506–6519, doi:10.1002/2014JD021542.

Lab Task 1: Suggested Approach

Using R, compare box plots of seasonal TC number and seasonal TC intensity to quantify and visualize the spread due to internal variability and model physics.

Download data and follow instructions under the 'Uncertainty' link under 'Lab Projects' tab on our website.

TC track data zip files:

- initial_condition_ensemble.zip
- physics_ensemble.zip

Lab Task 1: Results



Ensemble Climate Simulation



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Impact of Model Bias



Slingo and Palmer (2011)

Computational Limitations

A balance between competing multiplicative demands:



Computational Limitations



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Computational Limitations



Sufficient complexity to capture the extremes. Run ensemble and one member at high-resolution.

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Ensemble Considerations

- Which uncertainty to sample?
- Weight the members, even discard some?

When uncertainty poorly understood, use less quantitative measures: range or sign of a change (Kandlikar et al, 2005).

All ensembles are useful, the key is the interpretation.

More Uncertainty Sources!



Computational Uncertainty

Analysis Uncertainty



Be consistent across datasets

Asuka Suzuki-Parker



Observational Uncertainty

What we observe is not nature itself, but nature exposed to our method of questioning. – Heisenberg



Large impact of observed rainfall uncertainty on drought index



Differences among observational datasets can have similar magnitude to model error



Prein and Gobiet (2016)

Use multiple observational datasets for model evaluation

Computational Uncertainty

Experimental Design:





Computational Uncertainty



Differences in annual numbers and interannual variability

A Goal For Ensemble Simulation

The goal is not to reproduce the observed weather extremes, but rather to produce a range of possible scenarios consistent with known large-scale conditions.

Historical weather extremes are just one of many possibilities.



Relevant Uncertainty

Threshold sensitive decisions:



Focus uncertainty assessment on the scales and variables of interest

Useful Uncertainty: Example



Provide useful uncertainty

Prioritize your Uncertainty Sources



Laboratory Task 2

- It is May 1, 1998.
- A shipping insurer has exposure across the North Atlantic.
- The insurer is solvent up to 9 TCs.

Should the insurer buy protection against 9 or more TCs?

Lab Task 2: Suggested Approach

Using R, calculate and visualize the exceedance probability of 9 TCs, by fitting a distribution function to the seasonal TC numbers from the ensemble simulations.

Download data and follow instructions under the 'Uncertainty' link under the 'Lab Projects' tab on our website.

TC track data zip files

- initial_condition_ensemble.zip
- physics_ensemble.zip

Lab Task 2: Results

Histogram of ic_members.count





- Determining important sources of uncertainty for your project.
- Understanding uncertainty through ensemble modeling.
- Concept of relevant and useful uncertainty.







References

- Curry J.A, and Webster P.J. 2011: Climate science and the uncertainty monster. *Bull Am Meteor Soc* 92(12): 1667–1682.
- Deser, C., R. Knutti, S. Solomon, and A. S. Phillips, 2012: Communication of the role of natural variability in future North American climate. *Nat. Clim. Change*, 2, 775-779.
- Done, J.M., Bruyère, C.L., Ge, M and A. Jaye 2014: Internal Variability of North Atlantic Tropical Cyclones, J GR-Atmospheres. doi: 10.1002/2014JD021542
- Done, J.M., Holland, G.J., Bruyère, C.L., Leung, L.R., and Suzuki-Parker, A., 2013: Modeling high-impact weather and climate: Lessons from a tropical cyclone perspective. *Climatic Change.* doi: 10.1007/ s10584-013-0954-6
- Hawkins, Ed, Rowan Sutton, 2009: The Potential to Narrow Uncertainty in Regional Climate Predictions. *Bull. Amer. Meteor. Soc.*, 90, 1095–1107.
- Knutti, R., 2008: Should we believe model predictions of future climate change? *Philos. Trans. Roy. Soc. London,* A366, 4647–4664
- Prein, A. F. and Gobiet, A. 2016: Impacts of uncertainties in European gridded precipitation observations on regional climate analysis. Int. J. Climatol. doi:10.1002/joc.4706
- Slingo J, Palmer TN. 2011: Uncertainty in weather and climate prediction. *Phil. Trans. R. Soc. A* 369: 4751–4767.
- Sriver RL, CE Forest, Keller K 2015: Effects of initial conditions uncertainty on regional climate varia- bility: An analysis using a low-resolution CESM ensemble, Geophys. Res. Lett., 42, 5468–5476, doi: 10.1002/2015GL064546
- Stainforth, D. A., M. R. Allen, E. R. Tredger, and L. A. Smith, 2007: Confidence, uncertainty and decisionsupport relevance in climate predictions. *Philos. Trans. Roy. Soc. London,* A365, 2145–2161.
- Tye, M. R., G. Holland, and J. M. Done, 2014: Rethinking failure: Time for closer engineer-scientist collaborations on design. Proceedings of the Institution of Civil Engineers Forensic Engineering, 168, 49-57, doi:10.1680/feng.14.00004.