



Jet Propulsion Laboratory
California Institute of Technology

Probability as a Foundation for Data Uncertainty: Applications in Remote Sensing

Jon Hobbs ¹

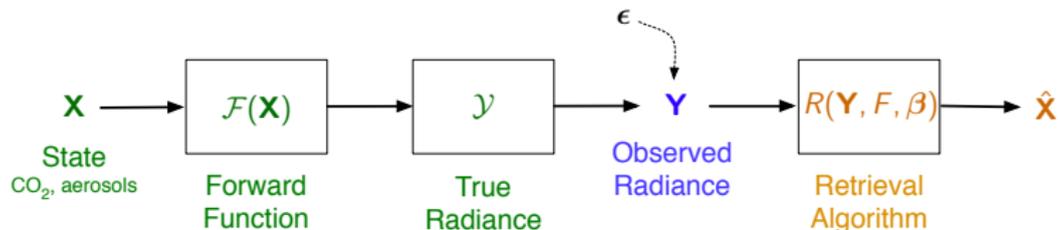
¹Jet Propulsion Laboratory, California Institute of Technology

Data Uncertainty

- Data uncertainty represents lack of knowledge about a geophysical quantity of interest (QOI) *after observing relevant data*.
- The true value of the QOI, \mathbf{X} , is generally unknown, so plausible/likely values must be characterized.
- Probability offers a coherent framework for representing the distribution of the QOI, or the plausible error $\hat{\mathbf{X}} - \mathbf{X}$, given an estimate $\hat{\mathbf{X}}$ based on observed data.
- Earth science data records are relying on increasingly complex methods for constructing estimates $\hat{\mathbf{X}}$.
 - Remote sensing retrievals using satellite radiances and radiative transfer models (Rodgers, 2000)
 - Data assimilation using Earth system models and multiple data sources

- National Research Council report (NRC, 2012) places uncertainty quantification (UQ) for complex physical systems in a probabilistic framework.
- UQ methodology seeks to identify the impact of sources, or contributors, to the distribution of the error for a quantity of interest (QoI).
- A probabilistic framework benefits from representing the system as a data-generating process, with the QoI as an outcome.
- Monitoring the process includes describing the prediction error under a particular set of conditions, such as a particular version of a retrieval algorithm.
- Improving the process can result from improved understanding of error sources.
- UQ has a role in both monitoring and improvement.

Observing System

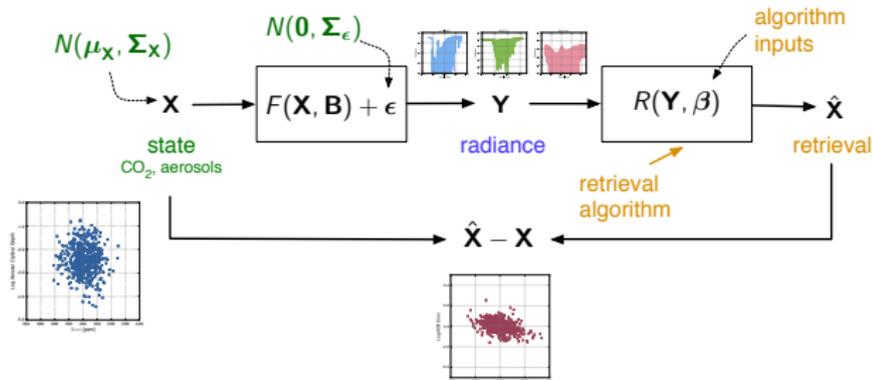


- Remote sensing observing system is a complex data-generating process with several key components.
 - True top-of-atmosphere radiance is a function of atmospheric state.
 - Instrument observes noisy radiance.
 - Retrieval algorithm produces estimate of state.
 - Science data system scales processing.
- Objective is inference on the state given the observed radiances, an *inverse problem*.

Approaches

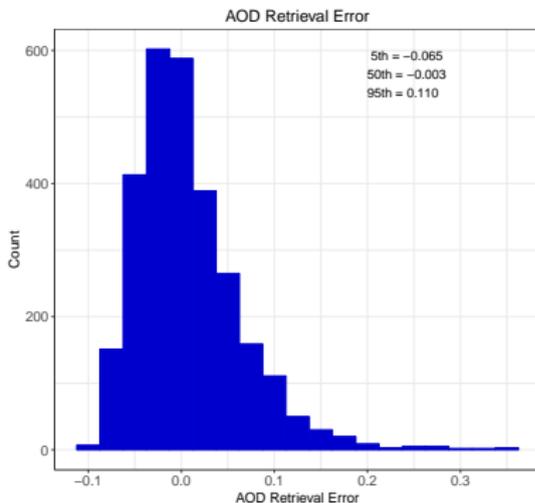
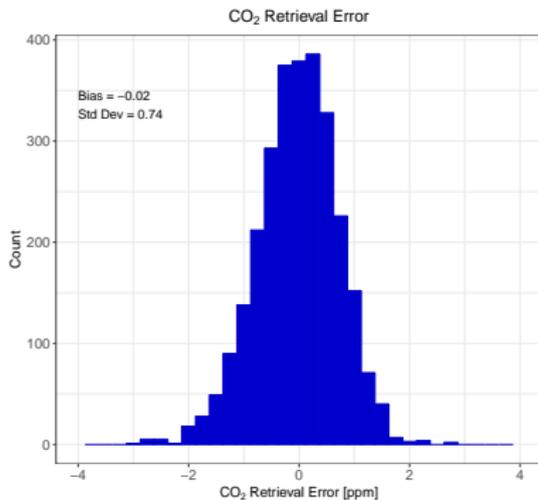
Multiple approaches for probabilistic assessment of observing systems

- In situ validation: Summarize the error distribution, $\hat{\mathbf{X}} - \mathbf{X}$, where substantially more accurate and precise observations of \mathbf{X} are available.
- Simulation studies: Monte Carlo experiments with the data-generating process, estimation procedure, and ensembles of user-specified true QOIs \mathbf{X} .



Error Distributions

- How should uncertainty be summarized?
 - Bias, variance may be sufficient for a symmetric error distribution.
 - Quantiles may be more appropriate for skewed, multi-modal distributions.



Current Work

- Toward Unified Error Reporting (TUNER): International effort to provide validation-based error assessment for retrievals of comparable QOIs from different satellites.
<http://www.issibern.ch/teams/tuner/>
- NASA AIST effort to develop tools for simulation-based UQ for retrievals (Hobbs et al., 2017)
 - Application to OCO-2 and AIRS Level 2 retrievals
- JPL internal initiative on UQ for Earth science applications

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Questions?

Jonathan.M.Hobbs@jpl.nasa.gov

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