

INTEGRATED FIELD RESEARCH CHALLENGE SITE Hanford 300 Area



Yoram Rubin¹, Xingyuan Chen¹, Haruko Murakami², Melanie S. Hahn¹, Mark L. Rockhold³, Vince R. Vermeul³, Glenn E. Hammond³, Peter C. Lichtner⁴

¹Department of Civil and Environmental Engineering, University of California, Berkeley, California, USA, CA, 94720

²Department of Nuclear Engineering, University of California, Berkeley, California, USA, CA, 94720



³Pacific Northwest National Laboratory, Richland, WA, 99352

⁴Los Alamos National Laboratory, Los Alamos, NM, 87545

Background

Our team focuses on development of the hydrologic model of the site through assimilating of the various types of data collected at the site, including data from boreholes and from a variety of tests and surveys.

Our efforts proceed along two complimentary lines of research. The first line is the development of a new approach for data inversion and assimilation called Method of Anchored Distributions (MAD, Rubin et al., 2009) and the numerical code needed for its implementation. The MAD framework is presented in Box 2.

The second line is the implementation of MAD to the Hanford IFRC site, which is a unique research site for studying and developing state-of-the-art inverse models because of the diversity of data types being acquired and the multiple experiments being conducted at the site. These two elements combined offer unprecedented opportunities for testing and improving MAD and answering fundamental questions related to 3D hydrogeological and geochemical characterization.

Our implementation efforts thus far focused on assimilating data from (1) the constant-rate injection tests, (2) the electromagnetic borehole flow meter (EBF) data, and (3) the March 2009 tracer test. Results from this effort are presented in Box 3. The implementation effort included linking MAD with the flow and transport code PFLOTRAN.

Theoretical Development

Our primary achievement on the theoretical side is the development and implementation of a new theory for stochastic inverse modeling and uncertainty quantification – MAD. MAD is built to assimilate multiple types of data that are collected at multiple observation scales, and that are related directly or indirectly to the target variables (e.g., conductivity, porosity). The data are assimilated using a Bayesian framework, and the end product is a services of 3D realizations of the conductivity field.

The significant points of departure of MAD from the current concepts are as follows: 1. It employs a theoretically sound method for moving information across multiple scales of





Assimilation of Experimental data at the IFRC Site

PFLOTRAN Integrated into MAD



Conditional Hydraulic Conductivity Fields

Results

Demonstrates high-low-high vertical profile along the





PFLOTRAN Conceptual Model

with the section of t

Tracer Breakthrough Curves



Technical Challenges and Near-Term Plans Technical issues:

- · Accurate specification of boundary conditions (e.g. interpolation between wells)
- · Coupled modeling of saturated and unsaturated zones.
- · Accurate modeling of vertical flows in wells.

Near-term plans:

- Assess optimal strategies for the likelihood function (e.g. which temporal moment or combination of moments of the breakthrough curves to use? OR time-series representation?)
- · Add porosity to the list of target variables.
- · Incorporate lithology information into MAD, possibly through the use of soft conditioning.

Future Research

We plan to pursue research questions related to 3D hydrogeological and geochemical characterization:

- How to improve site characterization using data from tracer experiments? Specifically, how to benefit from concentration measurements, as opposed to relying on aggregate measures of spatial spread?
- How to improve site characterization using data from geophysical surveys? Specifically, how to benefit from coupling data from resistivity surveys with hydrologic data in light of the large uncertainty associated with perophysical models and the difficulty in scaling them from the lab to the field?
- How to quantify uncertainty accurately, and how much uncertainty can be tolerated such that the research hypotheses of this project could be accepted with confidence?
- Given the high computational cost of 3D data interpretation, can we develop and implement a method for continuous updating of parameter estimates and uncertainty, building upon previous updates, as increasing amount of information become available?

References

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