

Model Analyses of Complex Systems Behavior using MADS

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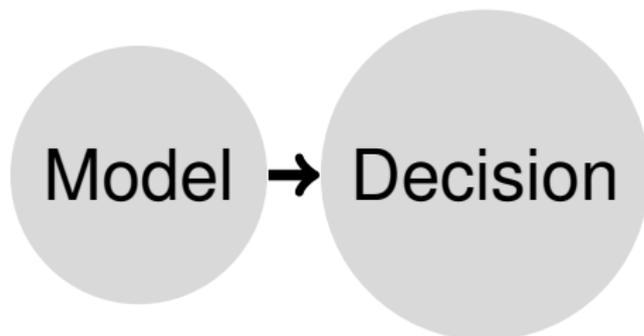
- ▶ **Climate Science**: Should we cap carbon emissions or not?
- ▶ **Meteorology**: Should we evacuate a city due to a hurricane?
- ▶ **Geology**: How much should we bid on a fossil fuel play?
- ▶ **Seismology**: Should we inject fluids in the underground (and how to do it without causing earthquakes and contamination)
- ▶ **Hydrogeology**: How to provide clean water supply?
- ▶ **Hydrogeology**: Which remediation option will clean up the groundwater?

Our work inform important decisions

- ▶ **Climate Science**: Should we cap carbon emissions or not?
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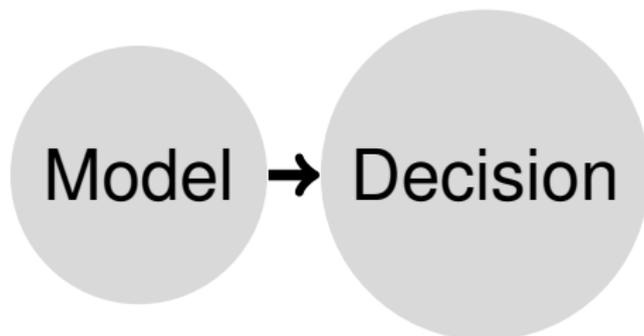
We rely on **data** & **models** to make scientifically defensible **decisions**

How should we support these decisions?



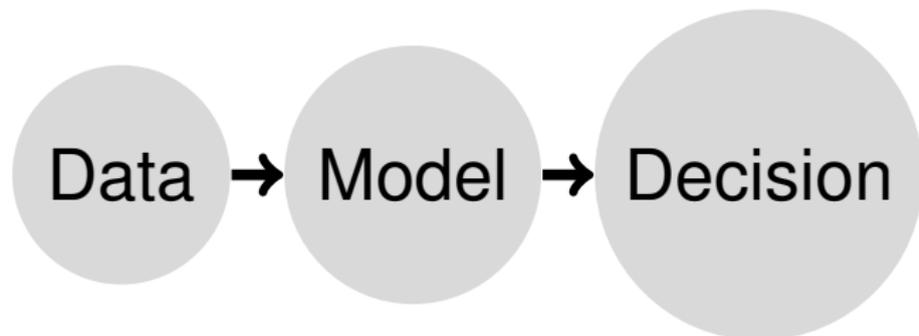
- ▶ Build a “representative” model
- ▶ Use the “representative” model to make a decision

How should we support these decisions?



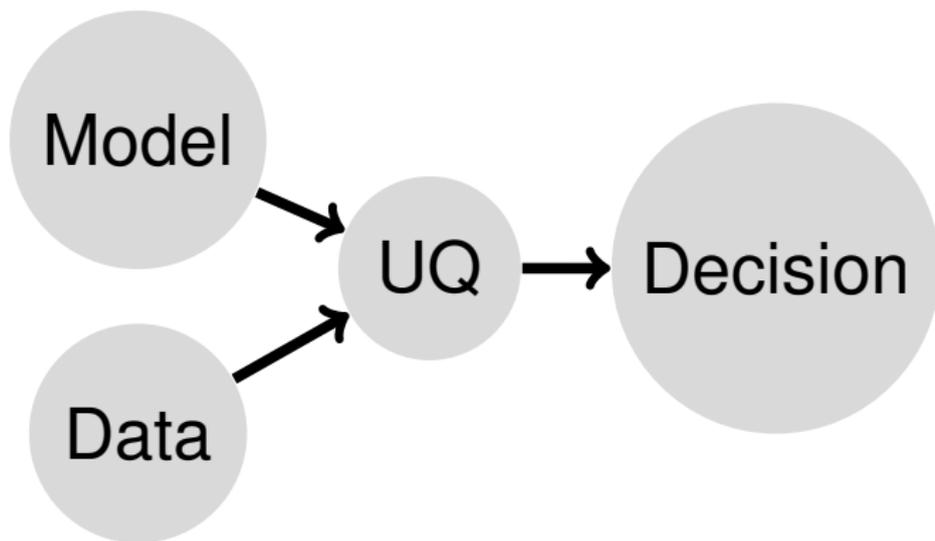
- ▶ Build a “representative” model
- ▶ Use the “representative” model to make a decision
- ▶ **However:**
 - ▶ many real-world models cannot be validated (especially in the earth sciences)
 - ▶ data can be highly uncertain
 - ▶ conceptualization can be highly uncertain
 - ▶ model predictions can be highly uncertain

How should we support these decisions?



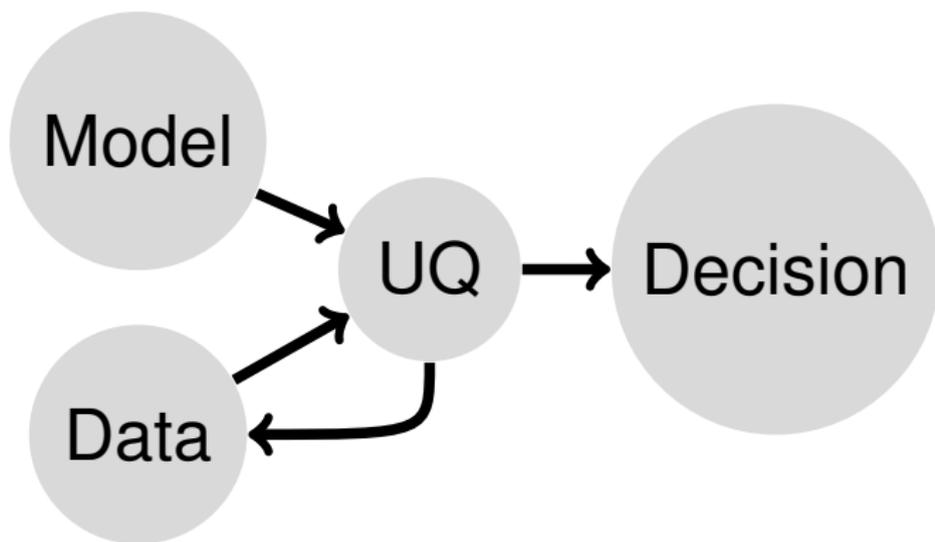
- ▶ Use data to calibrate the model
- ▶ Use the model to make a decision

How should we support these decisions?



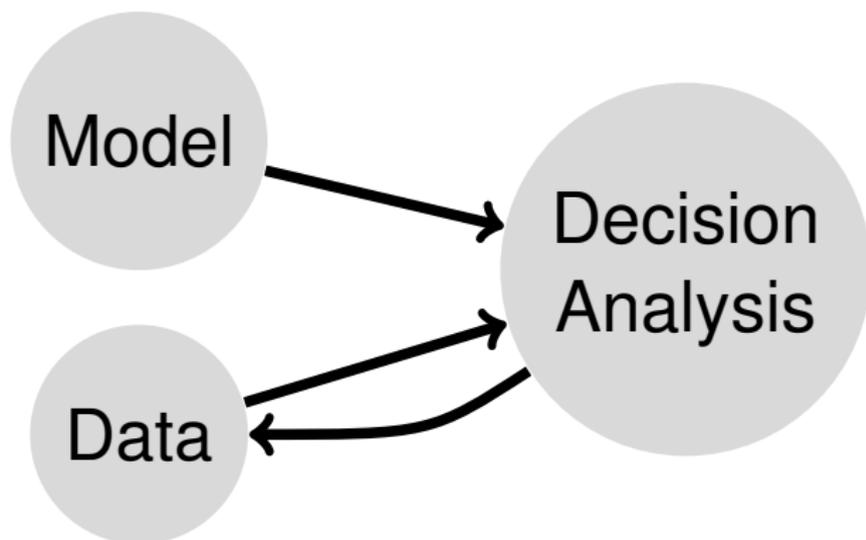
- ▶ Quantify uncertainty in the data and model
- ▶ Use estimated uncertainties in model predictions to make a decision

How should we support these decisions?



- ▶ Quantify uncertainty in the data and model
- ▶ Estimate uncertainties in model predictions
- ▶ Collect data that reduces the prediction uncertainties (optimal experimental design)
- ▶ Use the new data to quantify uncertainty again
- ▶ Use updated uncertainties to make decision (**are we done?**)

How should we support these decisions?



- ▶ Perform **decision analysis** coupling UQ with the decision process to evaluate uncertainty in decisions (**not** uncertainty in model parameters/predictions)
- ▶ Use the **decision analysis** to guide collection of data that can influence a better decision (it may not always be feasible)
- ▶ Use the new collected data to make a better decision

- ▶ We need robust and versatile decision analysis methodologies and tools
- ▶ Recently, we have developed a series of novel methods and techniques for **data**- and **model**-based **decision** analyses
- ▶ Most of them are implemented in **MADS**

Model Analysis & Decision Support



MADS: Model Analysis and Decision Support

- ▶ **MADS** is a **high-performance** computational framework
- ▶ **MADS** performs a wide range of **data-** & **model-**based analyses including
 - ▶ Sensitivity Analysis
 - ▶ Parameter Estimation, Model Inversion/Calibration
 - ▶ Uncertainty Quantification
 - ▶ Machine Learning Methods
 - ▶ Reduced Order Modeling (ROM)
 - ▶ Optimal Experimental Design (OED)
 - ▶ Decision Analysis
- ▶ **MADS** is open source code (GPL3) written in **julia**
- ▶ **julia** is a high-level, dynamic programming language for technical computing
- ▶ **julia** has C speed but with MatLab/Python flexibility
- ▶ **julia** provides access to a vast number of mathematical, statical, and visualization packages

MADS: Model Analysis and Decision Support

- ▶ **MADS** can be applied to perform analyses using any existing physics simulator
- ▶ **MADS** provides tools for model development, integration and coupling
- ▶ **MADS** utilizes advanced code development tools for
 - ▶ version control (git)
<https://github.com/madsjulia>
 - ▶ continuous integration (Travis-CI)
<https://travis-ci.org/madsjulia/Mads.jl>
 - ▶ tracking code test coverage (Coveralls)
<https://coveralls.io/github/madsjulia/Mads.jl>
- ▶ **MADS** contributors and developers are welcome
- ▶ **MADS** examples, manuals and publications are available at:
<https://mads.lanl.gov>
<https://madsjulia.github.io/Mads.jl>
<https://mads.readthedocs.io>

- ▶ **Information-Gap Decision Theory (IGDT)**
 - ▶ O'Malley, D., Vesselinov, V.V., Groundwater remediation using the information gap decision theory, Water Resources Research, doi: 10.1002/2013WR014718, 2014.
 - ▶ Harp, D.R., Vesselinov, V.V., Contaminant remediation decision analysis using information gap theory, Stochastic Environmental Research and Risk Assessment (SERRA), doi:10.1007/s00477-012-0573-1, 2012.
- ▶ **Bayesian-Information-Gap Decision Theory (BIG-DT)**
 - ▶ O'Malley, Vesselinov: Groundwater Remediation using Bayesian Information-Gap Decision Theory (**West 3024**, Thursday, 17:00 - 17:15, **H44E-05**)
 - ▶ Grasinger, M., O'Malley, D., Vesselinov, V.V., Karra, S., Decision Analysis for Robust CO2 Injection: Application of Bayesian-Information-Gap Decision Theory, International Journal of Greenhouse Gas Control, doi: 10.1016/j.ijggc.2016.02.017, 2016.
 - ▶ O'Malley, D., Vesselinov, V.V., Bayesian-Information-Gap decision theory with an application to CO2 sequestration, Water Resources Research, doi: 10.1002/2015WR017413, 2015.
 - ▶ O'Malley, D., Vesselinov, V.V., A combined probabilistic/non-probabilistic decision analysis for contaminant remediation, Journal on Uncertainty Quantification, SIAM/ASA, doi: 10.1137/140965132, 2014.
- ▶ **Optimal Experimental Design (OED) driven by decision analysis**
 - ▶ O'Malley, D., Vesselinov, V.V., (in preparation).
- ▶ **Measure-theoretic Uncertainty Quantification (UQ)**
 - ▶ Dawson, Butler, Mattis, Westerink, Vesselinov, Estep: Parameter Estimation for Geoscience Applications Using a Measure-Theoretic Approach (**West 3024**, Thursday, 17:30 - 17:45, **H44E-07**)
 - ▶ Mattis, S.A., Butler, T.D. Dawson, C.N., Estep, D., Vesselinov, V.V., Parameter estimation and prediction for groundwater contamination based on measure theory, Water Resources Research, doi: 10.1002/2015WR017295, 2015.
- ▶ **Novel Levenberg-Marquardt (LM) optimization method using a dimensionality reduction based on Krylov subspace method**
 - ▶ Lin, Y, O'Malley, D., Vesselinov, V.V., A computationally efficient parallel Levenberg-Marquardt algorithm for highly parameterized inverse model analyses, Water Resources Research, doi: 10.1002/2016WR019028, 2016.

- ▶ Model inversion using modified Total-Variation (**TV**) regularization
 - ▶ Lin, O'Malley, Vesselinov: Hydraulic Inverse Modeling with Modified Total-Variation Regularization with Relaxed Variable-Splitting (**poster**, Thursday, 8:00 - 12:00, **H41B-1301**)
- ▶ Model inversion using Principal Component Geostatistical Approach (**PCGA**) and Randomized Geostatistical Approach (**RG**)
 - ▶ Lin, Y, Le, E.B, O'Malley, D., Vesselinov, V.V., Bui-Thanh, T., Large-Scale Inverse Model Analyses Employing Fast Randomized Data Reduction, 2016, (submitted).
- ▶ Blind Source Separation (**BSS**) using Non-negative Matrix Factorization (**NMF**)
 - ▶ Vesselinov, V.V., O'Malley, D., Alexandrov, B.S., Source identification of groundwater contamination sources and groundwater types using semi-supervised machine learning, (in preparation).
 - ▶ Iliev, F.L., Stanev, V.G., Vesselinov, V.V., Alexandrov, B.S., Sources identification using shifted non-negative matrix factorization combined with semi-supervised clustering, 2016, (submitted).
 - ▶ Stanev, V.G., Iliev, F.L., Vesselinov, V.V., Alexandrov, B.S., Machine learning approach for identification of release sources in advection-diffusion systems, 2016, (submitted).
 - ▶ Alexandrov, B., Vesselinov, V.V., Blind source separation for groundwater level analysis based on non-negative matrix factorization, Water Resources Research, doi: 10.1002/2013WR015037, 2014.
- ▶ Support Vector Regression (**SVR**) methods for surrogate modeling
 - ▶ Alexandrov, B.S., O'Malley, D., Vesselinov, V.V., (in preparation).
 - ▶ Vesselinov, O'Malley, Alexandrov, Moore: Reduced Order Models for Decision Analysis and Upscaling of Aquifer Heterogeneity (**South 302**, Monday, 8:45 - 9:00, **NG11A-04**)

- ▶ Advanced Monte Carlo Methods: Robust Adaptive Metropolis (**RAM**) and Affine Invariant Markov Chain Monte Carlo Ensemble Sampler (aka **Emcee**)
 - ▶ Vihola: Robust adaptive Metropolis algorithm with coerced acceptance rate, Statistics and Computing, 2012.
 - ▶ Goodman, Weare: Ensemble samplers with affine invariance. Communications in applied mathematics and computational science, 2010.
- ▶ Extended Fourier Amplitude Sensitivity Testing (**eFAST**) global sensitivity analysis
 - ▶ Saltelli, et al. Global sensitivity analysis, John Wiley & Sons, 2008.
- ▶ Multifidelity Global Sensitivity Analysis (**MFSA**) under given computational budget
 - ▶ Qian, Peherstorfer, O'Malley, Vesselinov, Wilcox: Multifidelity Global Sensitivity Analysis, SIAM, 2016, (submitted).

▶ Groundwater contaminant remediation (LANL Chromium & RDX)

- ▶ Mattis, S.A., Butler, T.D. Dawson, C.N., Estep, D., Vesselinov, V.V., Parameter estimation and prediction for groundwater contamination based on measure theory, Water Resources Research, doi: 10.1002/2015WR017295, 2015.
- ▶ Vesselinov, V.V., O'Malley, D., Katzman, D., Model-Assisted Decision Analyses Related to a Chromium Plume at Los Alamos National Laboratory, Waste Management, 2015.
- ▶ O'Malley, D., Vesselinov, V.V., A combined probabilistic/non-probabilistic decision analysis for contaminant remediation, Journal on Uncertainty Quantification, SIAM/ASA, doi: 10.1137/140965132, 2014.
- ▶ O'Malley, D., Vesselinov, V.V., Analytical solutions for anomalous dispersion transport, Advances in Water Resources, doi: 10.1016/j.advwatres.2014.02.006, 2014.

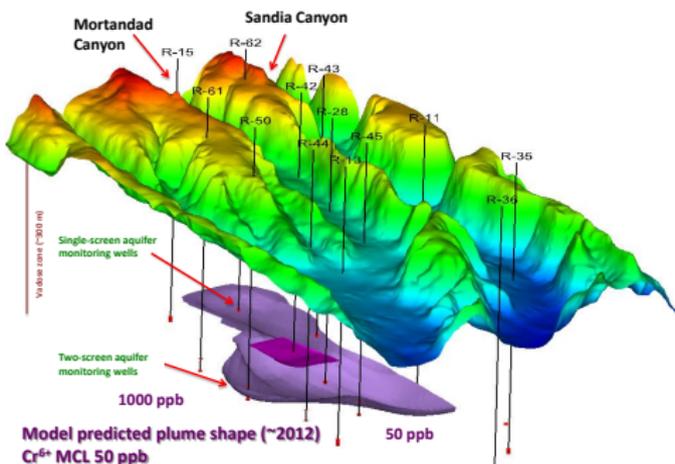
▶ Water/Energy/Food Nexus

- ▶ Zhang, Vesselinov: Bi-Level Decision Making for Supporting Energy and Water Nexus (**West 3016**: Wednesday, 09:15 - 09:30, **H31J-06**)
- ▶ Zhang, X., Vesselinov, V.V., Integrated Modeling Approach for Optimal Management of Water, Energy and Food Security Nexus Advances in Water Resources, Advances in Water Resources, 2016 (submitted).
- ▶ Zhang, X., Vesselinov, V.V., Energy-Water Nexus: Balancing the Tradeoffs between Two-Level Decision Makers Applied Energy, Applied Energy, DOI: 10.1016/j.apenergy.2016.08.156, 2016.

▶ CO₂ injection

- ▶ Grasinger, M., O'Malley, D., Vesselinov, V.V., Karra, S., Decision Analysis for Robust CO₂ Injection: Application of Bayesian-Information-Gap Decision Theory, International Journal of Greenhouse Gas Control, doi: 10.1016/j.ijggc.2016.02.017, 2016.
- ▶ O'Malley, D., Vesselinov, V.V., Bayesian-Information-Gap decision theory with an application to CO₂ sequestration, Water Resources Research, doi: 10.1002/2015WR017413, 2015.

LANL Chromium site



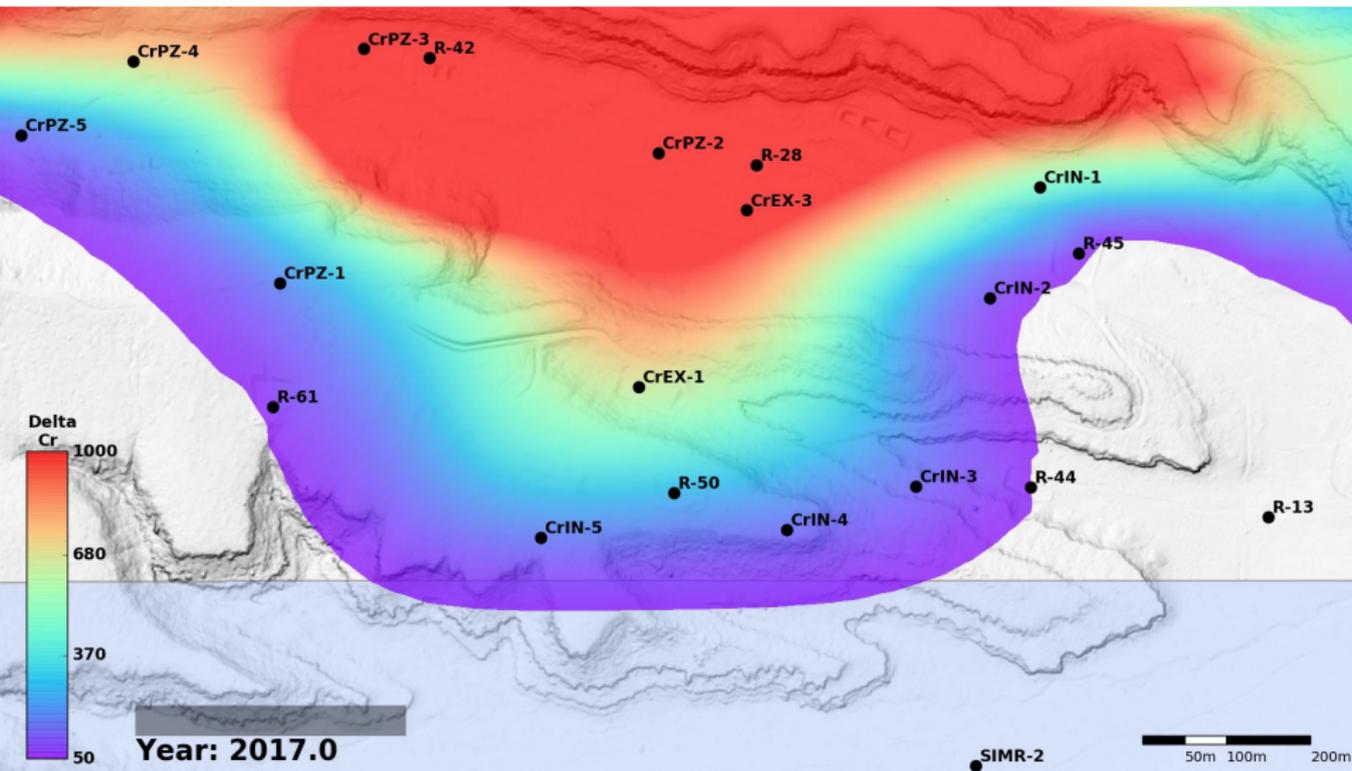
- ▶ Groundwater contamination site with high visibility (DOE)
- ▶ More than 20 wells drilled since 2007 (each wells costs \$2-3M)
- ▶ **Limited remedial options**

- ▶ Complex **uncertainties/unknowns**
- ▶ Plume is located near **LANL boundary** and **water-supply wells**
- ▶ Modeling accounts for complex biogeochemical processes in highly heterogeneous media
- ▶ In the last 5 years, we have accumulated close to 2,000 years computational time on the LANL HPC clusters for various model analyses
- ▶ Used up to 4,096 processors simultaneously

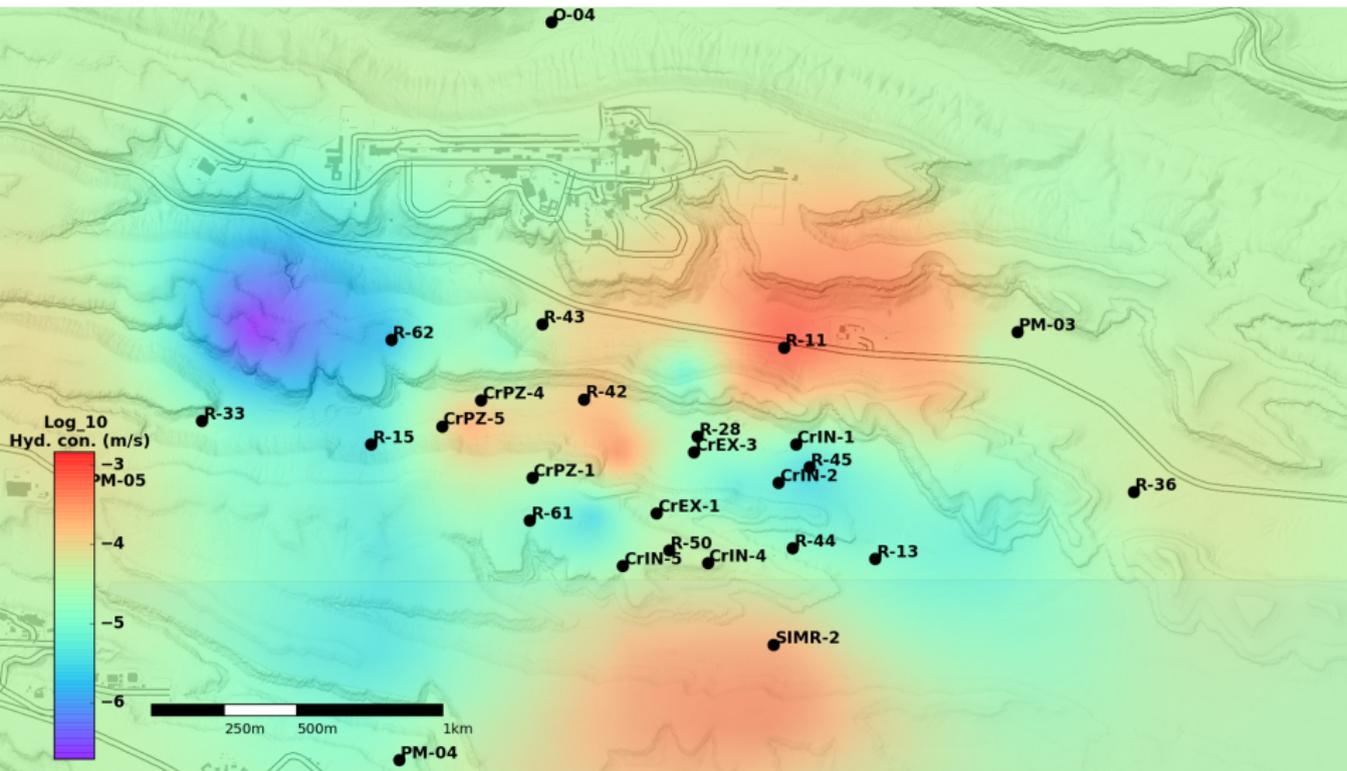
LANL regional aquifer model

- ▶ Model domain encompasses the regional aquifer beneath LANL ($\approx 8 \times 4 \times 0.3 \text{ km}$)
- ▶ 766,283 nodes / 4,659,062 cells
- ▶ 193 concentration calibration targets (representing annual transients for about 10 years)
- ▶ 182,090 water-level calibration targets (representing daily transients for about 4 years)
- ▶ water-level transients represent pumping effects caused by 9 wells (6 water-supply wells and 3 site wells where pumping tests are conducted)
- ▶ 230 unknown model parameters representing groundwater flow and transport (including aquifer heterogeneity and spatial location/strength of 3 unknown contaminant sources)
- ▶ Calibration required about 100 years computational time
- ▶ ... more data and physics/biogeochemistry needs to be incorporated in the model soon!

LANL regional aquifer model: Chromium transients



LANL regional aquifer model: Aquifer heterogeneity





LANL ADEM: Los Alamos National Laboratory Environmental Management Directorate



DiaMonD: An Integrated Multifaceted Approach to Mathematics at the Interfaces of Data, Models, and Decisions



`https://mads.lanl.gov`

`https://github.com/madsjulia`

`https://madsjulia.github.io/Mads.jl`



Related model and decision analyses presentations at AGU 2016

- ▶ Vesselinov, O'Malley, Alexandrov, Moore: Reduced Order Models for Decision Analysis and Upscaling of Aquifer Heterogeneity (**South 302**, Monday, 8:45 - 9:00, **NG11A-04**, **invited**)
- ▶ Lu, Vesselinov, Lei: Identifying Aquifer Heterogeneities using the Level Set Method (**poster**, Wednesday, 8:00 - 12:00, **H31F-1462**)
- ▶ Zhang, Vesselinov: Bi-Level Decision Making for Supporting Energy and Water Nexus (**West 3016**: Wednesday, 09:15 - 09:30, **H31J-06**)
- ▶ Vesselinov, O'Malley: Model Analysis of Complex Systems Behavior using MADS (**West 3024**: Wednesday, 15:06 - 15:18, **H33Q-08**)
- ▶ Hansen, Vesselinov: Analysis of hydrologic time series reconstruction uncertainty due to inverse model inadequacy using Laguerre expansion method (**West 3024**: Wednesday, 16:30 - 16:45, **H34E-03**)
- ▶ Lin, O'Malley, Vesselinov: Hydraulic Inverse Modeling with Modified Total-Variation Regularization with Relaxed Variable-Splitting (**poster**, Thursday, 8:00 - 12:00, **H41B-1301**)
- ▶ Hansen, Haslauer, Cirpka, Vesselinov: Prediction of Breakthrough Curves for Conservative and Reactive Transport from the Structural Parameters of Highly Heterogeneous Media (**West 3014**, Thursday, 14:25 - 14:40, **H43N-04**)
- ▶ O'Malley, Vesselinov: Groundwater Remediation using Bayesian Information-Gap Decision Theory (**West 3024**, Thursday, 17:00 - 17:15, **H44E-05**)
- ▶ Dawson, Butler, Mattis, Westerink, Vesselinov, Estep: Parameter Estimation for Geoscience Applications Using a Measure-Theoretic Approach (**West 3024**, Thursday, 17:30 - 17:45, **H44E-07**)