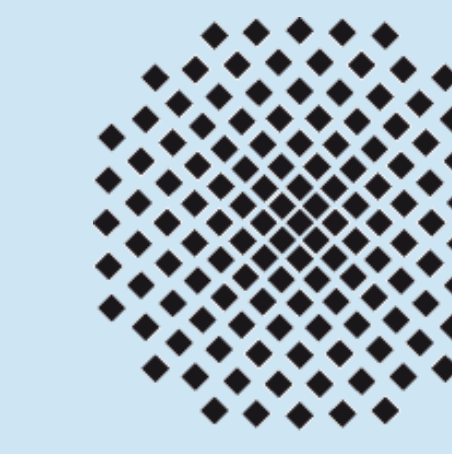
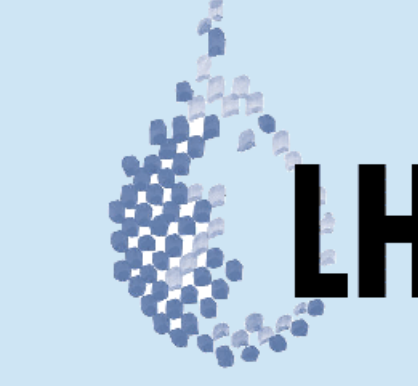


AN INTEGRATIVE DATA-ADAPTIVE APPROACH FOR GLOBAL SENSITIVITY ANALYSIS

APPLICATION TO SUBSURFACE FLOW AND TRANSPORT



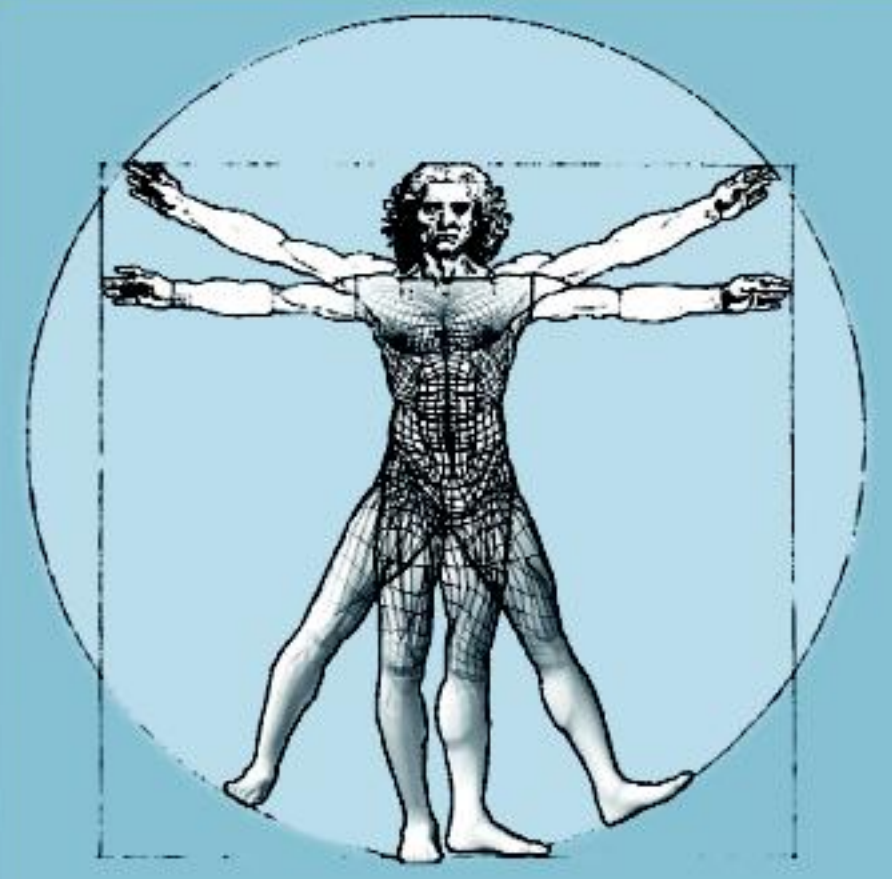
Universität Stuttgart
Germany



SimTech
Cluster of Excellence

FELIPE P. J. DE BARROS^{1,3}, SERGEY OLADYSHKIN^{1,2} & WOLFGANG NOWAK^{1,2}

¹ SRC SIMULATION TECHNOLOGY, ² INSTITUTE OF HYDRAULIC ENGINEERING, ³ INSTITUTE FOR APPLIED ANALYSIS AND NUMERICAL SIMULATION, UNIVERSITY OF STUTTGART, STUTTGART, GERMANY



MOTIVATION

Modeling uncertain systems continues to pose high demands. For example predicting flow and transport processes in the subsurface is a challenge since uncertainty in hydraulic properties of the subsurface is ubiquitous. Data sets are limited and costly. For such reasons, stochastic tools are required to support engineering design tasks under uncertainty, and sensitivity analysis with respect to uncertain model parameters yields valuable information.

OBJECTIVES

- Tackle global sensitivity analysis (GSA) and uncertainty quantification using aPC [1].
- Make use of a hybrid-analytical framework towards efficient computations.
- Provide method that takes into account different sources of information (raw data, statistical distributions, etc).

PHYSICAL SCENARIO

We will demonstrate our methodology for a contaminant transport problem in a 3D heterogeneous porous media. A solute is instantaneously released and undergoes purely advective transport. The aquifer has a hydraulic conductivity tensor and effective porosity.

For illustration purposes, we consider flow to be incompressible, single-phased, at steady-state, free of boundary effects and with velocity satisfying Darcy law:

$$\mathbf{u}(\mathbf{x}) = -\frac{\mathbf{K}(\mathbf{x})}{n_e(\mathbf{x})} \nabla h$$

where the hydraulic head is determined from the continuity equation:

$$\nabla \cdot [\mathbf{K}(\mathbf{x}) \nabla h(\mathbf{x})] = 0.$$

A tracer with initial concentration is instantaneously released from a rectangular source volume under purely advective transport conditions. Under these conditions, the governing equation for transport is:

$$\frac{\partial C}{\partial t} + u_i \frac{\partial C}{\partial x_i} = 0$$

With the aid of the Lagrangian framework, concentration mean can be expressed as [6]:

$$\langle C(\mathbf{x}, t) \rangle = C_0 \prod_{i=1}^3 \psi_i(\mathbf{x}, t) \quad \psi_i(\mathbf{x}, t) = \frac{1}{2} \operatorname{erf} \left[\frac{x_i - U_{it} + L_i/2}{\sqrt{2X_{ii}(t; \mathbf{a})}} \right] - \frac{1}{2} \operatorname{erf} \left[\frac{x_i - U_{it} - L_i/2}{\sqrt{2X_{ii}(t; \mathbf{a})}} \right]$$

REFERENCES

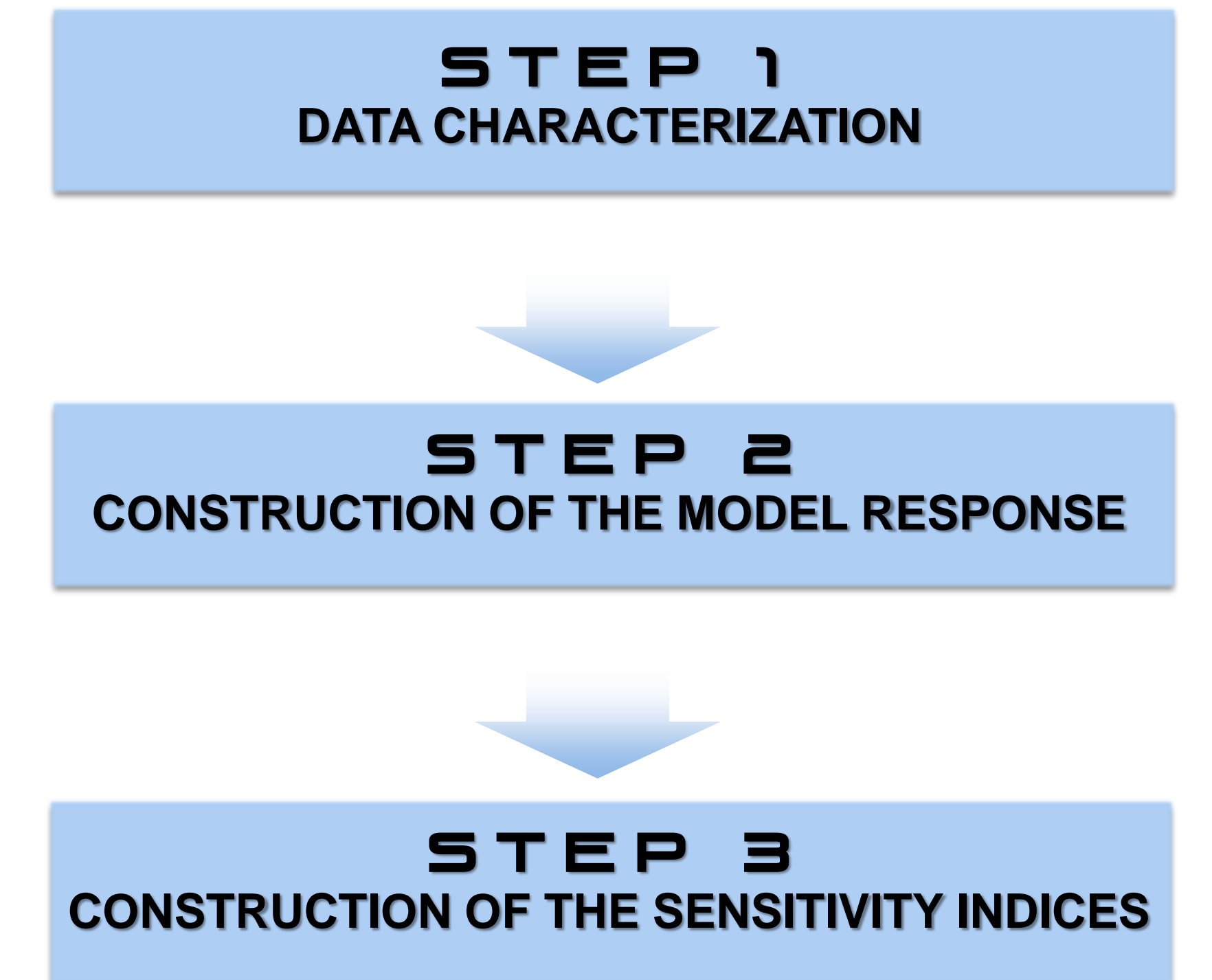
- [1] S. Oladyshkin, W. Nowak, Data-driven uncertainty quantification using the arbitrary polynomial chaos expansion, Reliability Engineering and System Safety. Submitted. 2010
- [2] I. M. Sobol, On sensitivity estimation for nonlinear mathematical models, Mathem. Mod. 2 (1) (1990) 112–118.
- [3] S. A. Homma, T. Importance measures in global sensitivity analysis of nonlinear models, Reliability Engineering and System Safety 52 (1) (1996) 1–17.
- [4] S. Oladyshkin, F.P.J. de Barros, W. Nowak, Global Sensitivity Analysis: Hybrid Data-Adaptive Framework and an Example from Heterogeneous Subsurface Flow and Transport. Advances in Water Resources. Submitted. 2011.
- [5] T. Crestaux, O. Le Maître, J.-M. Martinez, Polynomial chaos expansion for sensitivity analysis, Reliability Engineering and System Safety 94 (7) (2009) 1161–1172.
- [6] Y. Rubin, M. A. Cushey, A. Bellin, Modeling of transport in groundwater for environmental risk assessment, Stochastic Hydrol. Hydraul. 8 (1) (1994) 57–77.

METHODOLOGY

We propose a response surface method for global sensitivity analysis (based on the arbitrary Polynomial Chaos Expansion, aPC). We use analytical and hybrid analytical-numerical formulations that further improve computational efficiency.

The key advantage are:

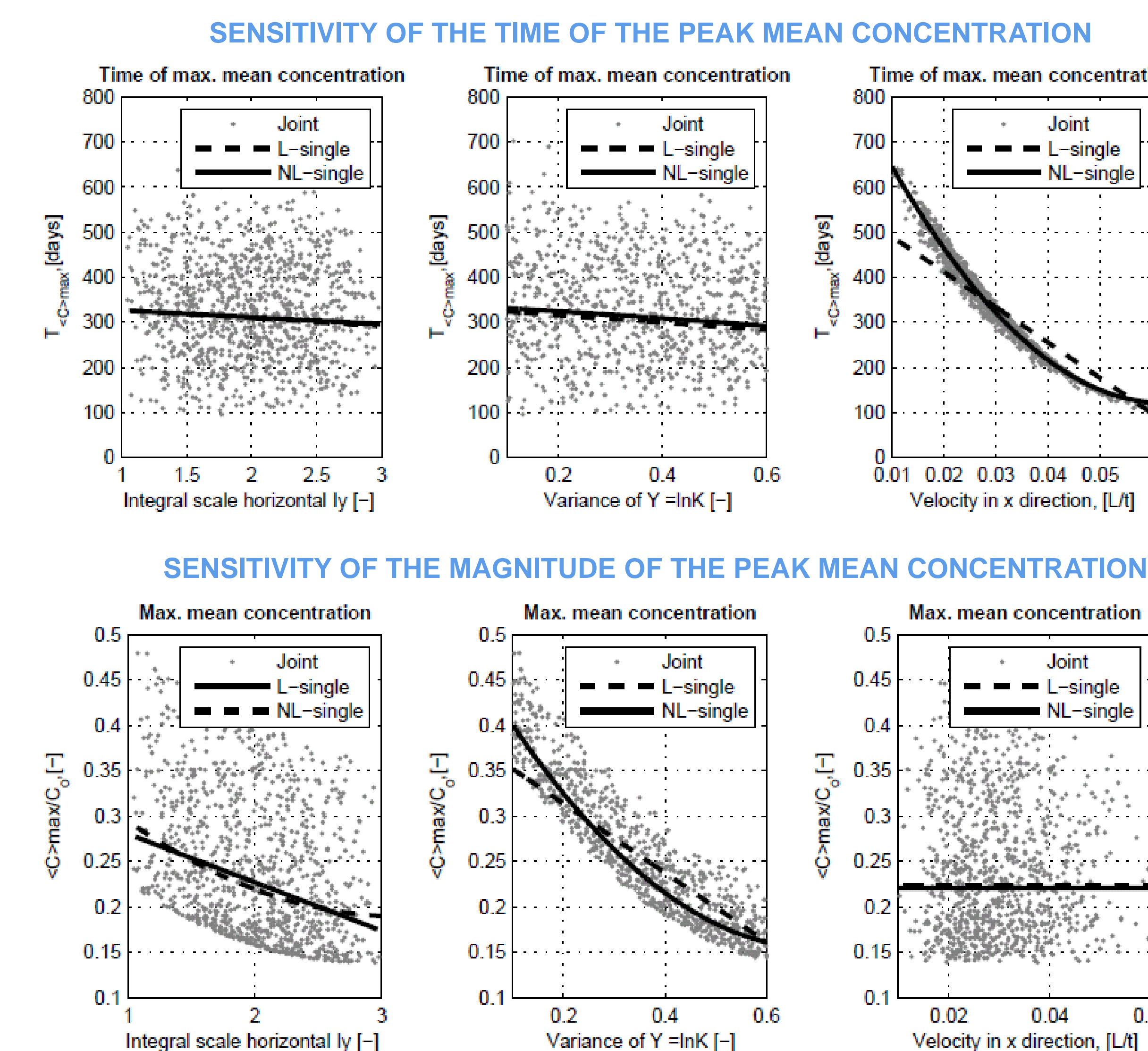
- reduces the computational burden associated with Monte Carlo methods or with conventional global sensitivity analysis that also requires many evaluations of a simulation model [5];
- method incorporates probabilities or weighting functions for the investigated model parameter and the full range of possible simultaneous outcomes;
- data-adaptive framework allows to incorporate this information while requiring only a finite number of statistical moments for the investigated parameters.



THREE-STEP GSA ALGORITHM

SENSITIVITY ANALYSIS

We perform a GSA for two distinct prediction: (1) time of peak mean concentration and (2) the magnitude of the maximum mean concentration. We study the sensitivity towards three uncertain parameters (integral scale, variance of logconductivity and mean longitudinal velocity).



QUANTITATIVE INFORMATION

Sobol Index [2] that indicates what fraction of the of total variance of output can be traced back to the joint contributions of the parameters $\{i, j, \dots, k\}$.

Total Index [3] expresses the total contribution to the variance of model output due to the uncertainty of an individual parameter

Weighted Sensitivity Index [4] reflects the slope due to an individual parameter, but averaged over all statistical distribution of all other parameters

Sobol index	S_1	S_2	S_3	S_{12}	S_{13}	S_{23}
Value for $T_{C, \max}$	0.004	0.016	0.975	8×10^{-4}	6×10^{-4}	0.003
Rank for $T_{C, \max}$	3	2	1	6	5	4
Value for C_{\max}	0.111	0.887	6×10^{-9}	0.002	9×10^{-10}	1×10^{-10}
Rank for C_{\max}	2	1	4	3	5	6

Total sensitivity index	S_1^T	S_2^T	S_3^T
Value for $T_{C, \max}$	0.005	0.021	0.980
Rank for $T_{C, \max}$	3	2	1
Value for C_{\max}	0.2113	0.889	8×10^{-9}
Rank for C_{\max}	2	1	3

Weighted sensitivity index	S_{w1}	S_{w2}	S_{w3}
Value for $T_{C, \max}$	14.246	66.163	2×10^4
Rank for $T_{C, \max}$	3	2	1
Value for C_{\max}	0.143	0.958	7×10^{-4}
Rank for C_{\max}	2	1	3

ACKNOWLEDGMENTS

The authors would like to thank the German Research Foundation (DFG) for financial support of the project within the Cluster of Excellence in Simulation Technology (EXC 310/1) at the University of Stuttgart.

