

A Multi-method Generalized Approach to Assess Sensitivity of Complex Watershed Models

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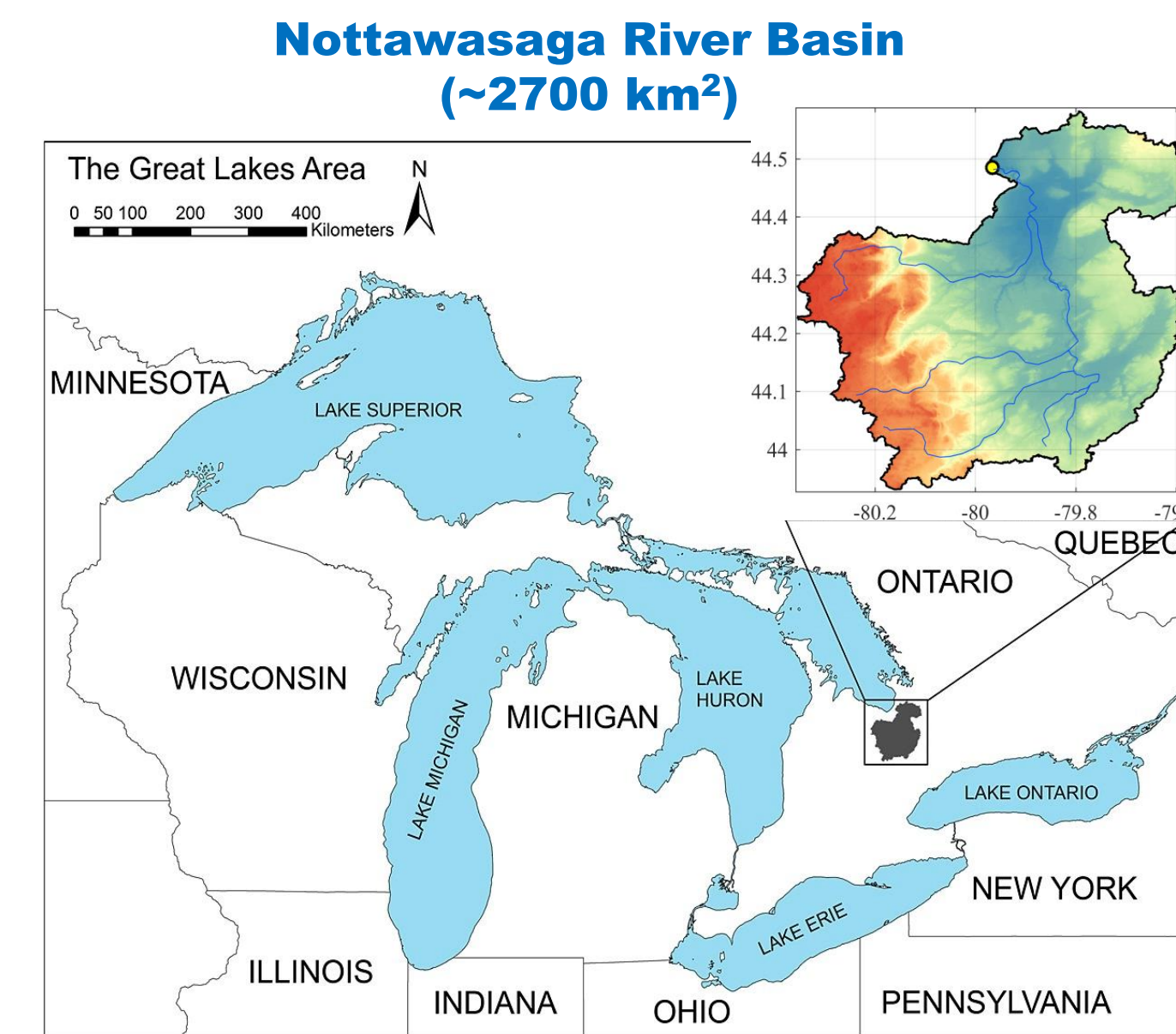
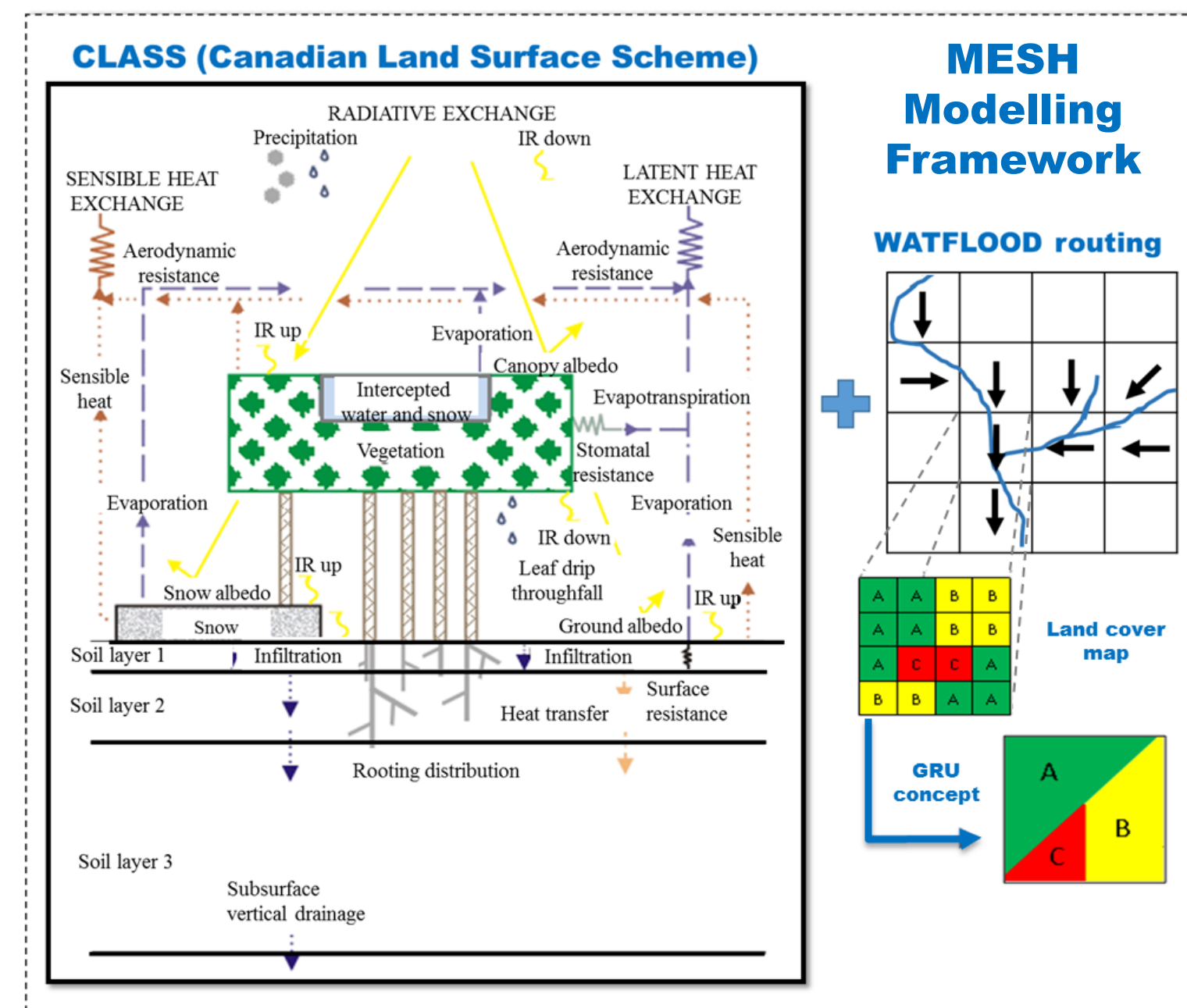
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INTRODUCTION

Dynamical physically-based watershed models are being increasingly used as the primary tool for water resources planning and management due to advances in computational power and data availability. For an enhanced and efficient development and application of these complex models, it is critical to understand the dynamical behavior of these models and identify the most influential factors (e.g., parameters) controlling it. Global Sensitivity Analysis (GSA) techniques can be used for this purpose.

The challenge is that GSA results depend on the GSA approach (e.g., derivative-based, variance-based, or variogram-based), and the type of model response considered. They can also vary with time. To address these challenges a new approach called Generalized Global Sensitivity Matrix (GGSM) is proposed. When coupled with STAR-VARS algorithm, GGSM, can use any GSA approach, and model response, and time-aggregated or time-varying sensitivity indices, to conduct a comprehensive GSA, and produce a wealth of model sensitivity information, with only one single GSA experiment.



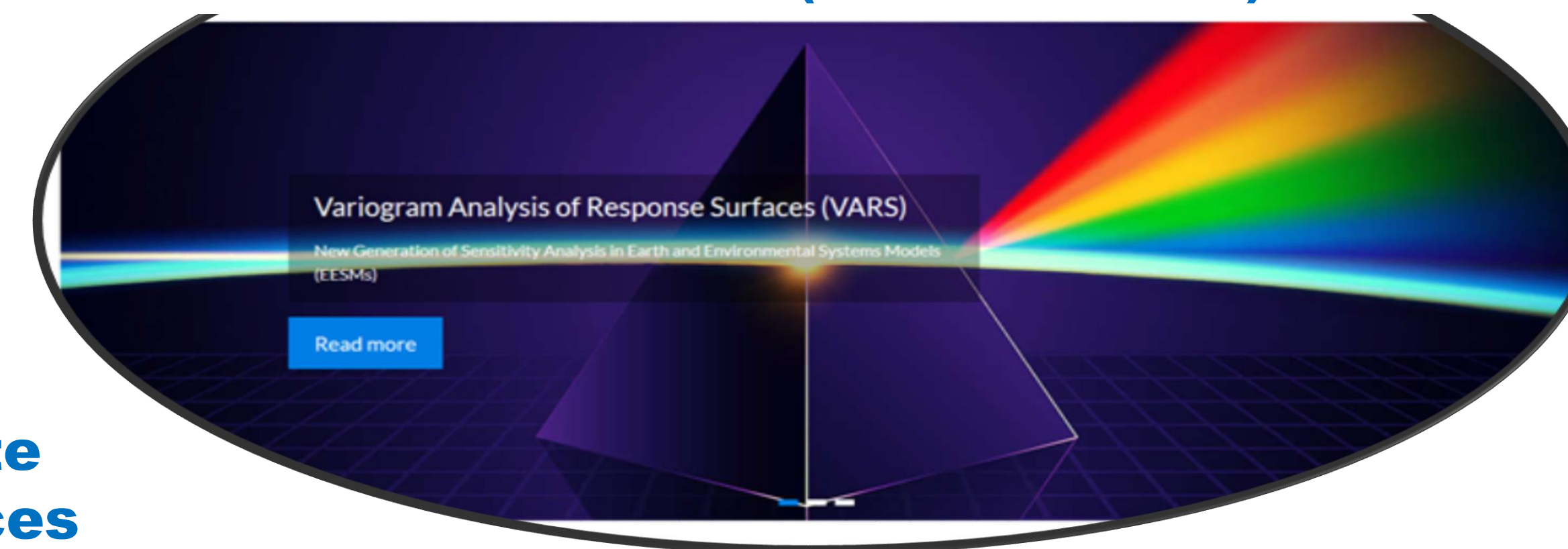
31 parameters for Crop GRU

Parameter	Description	Range
LAI05	Annual maximum leaf area index	(2.0, 4.0)
LAI06	Natural logarithm of the roughness length	(-2.0, -1.0)
ALIC	Average visible albedo when fully-leaved	(0.04, 0.08)
ALIC	Average near-infrared albedo when fully-leaved	(0.2, 0.3)
CMAS	Annual maximum canopy mass (kg m ⁻²)	(2, 5)
SCUT	Annual maximum rooting (m)	(0.2, 0.4)
SCUT	Minimum stomatal resistance (s m ⁻¹)	(60, 100)
CLAI0	Reference value of shortwave radiation (W m ⁻²)	(10, 50)
VZ05	Vapour pressure deficit coefficient 'W'	(0.2, 0.8)
VZ06	Vapour pressure deficit coefficient 'W'	(0.7, 1.3)
PSGA	Soil moisture suction coefficient 'W'	(75, 125)
ZS05	Soil moisture suction coefficient 'W'	(2, 8)
SANDP	Percent sand in the mineral soil of layer 1 (%)	(85, 100) sandy soil
CLAI1	Percent clay in the mineral soil of layer 1 (%)	(0, 10) clay loam & shy clay loam
ORGI	Percent organic matter in the mineral soil of layer 1 (%)	(0, 5)
SDEP	Soil permeability (depth) (m)	(0.1, 4.0)
THL	Threshold depth above which snow coverage is considered 100% (m)	(0.05, 0.2)
ZPLS	Maximum water ponding depth for snow-covered areas (m)	(0.02, 0.15)
ZPLS	Maximum water ponding depth for snow-free areas (m)	(0.02, 0.15)
SDEN	Drainage density (km km ⁻²)	(2, 100)
XSLP	Estimated average slope of the GRU type	(0.0001, 0.04)
ORGI	Ratio of saturated horizontal hydraulic conductivity at a depth of 1 meter to the saturated horizontal hydraulic conductivity at the surface	(0.0001, 1)
MANN	Manning's roughness coefficient 'n'	(0.02, 2)
RATIO	The ratio of horizontal to vertical saturated hydraulic conductivity	(2, 100)
WTR2	Channel roughness factor	(0.02, 2)

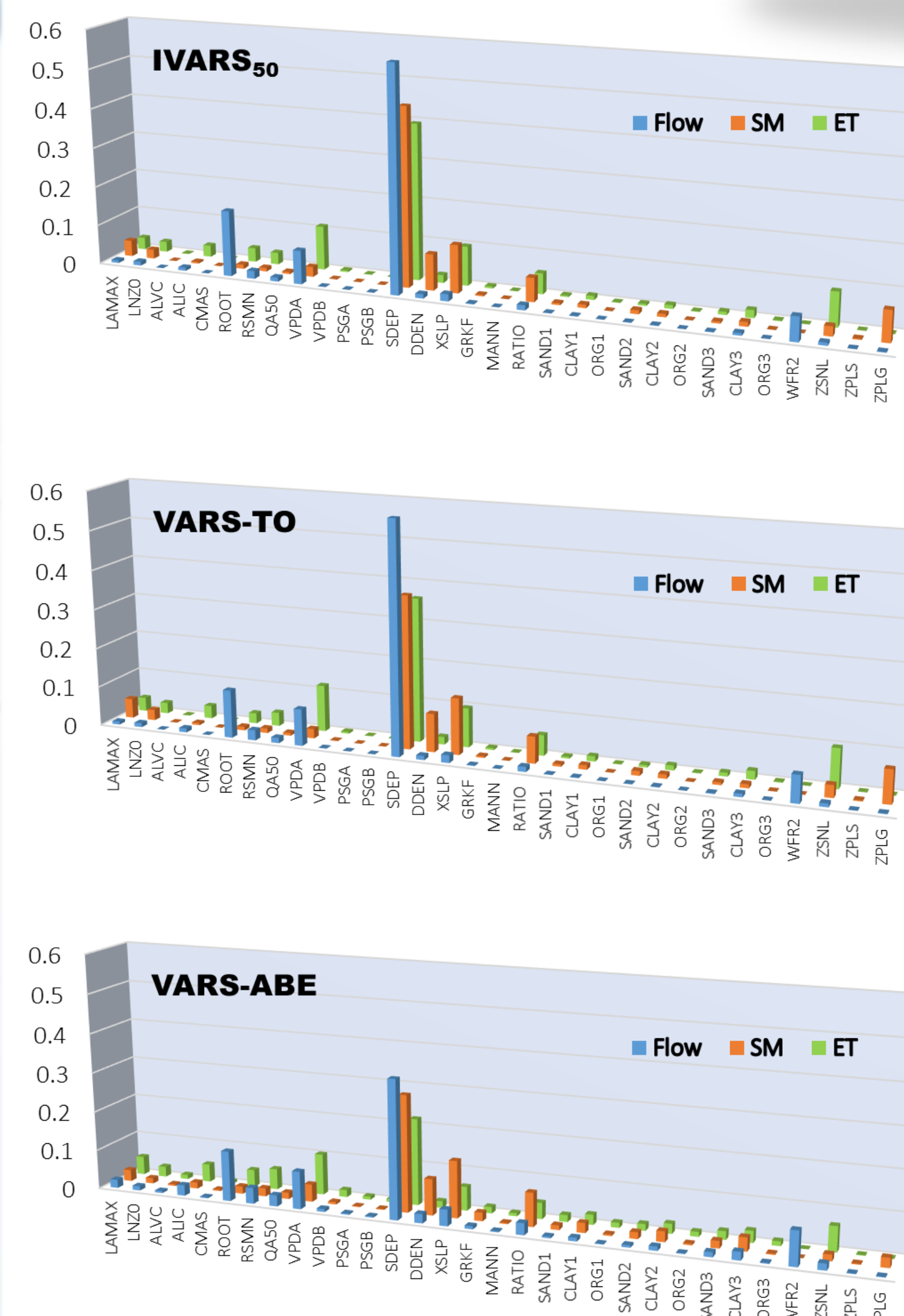
VARS information

- 100 STARS
- 31 Parameters
- VARS resolution = 0.1
- Total number of model runs = 28000

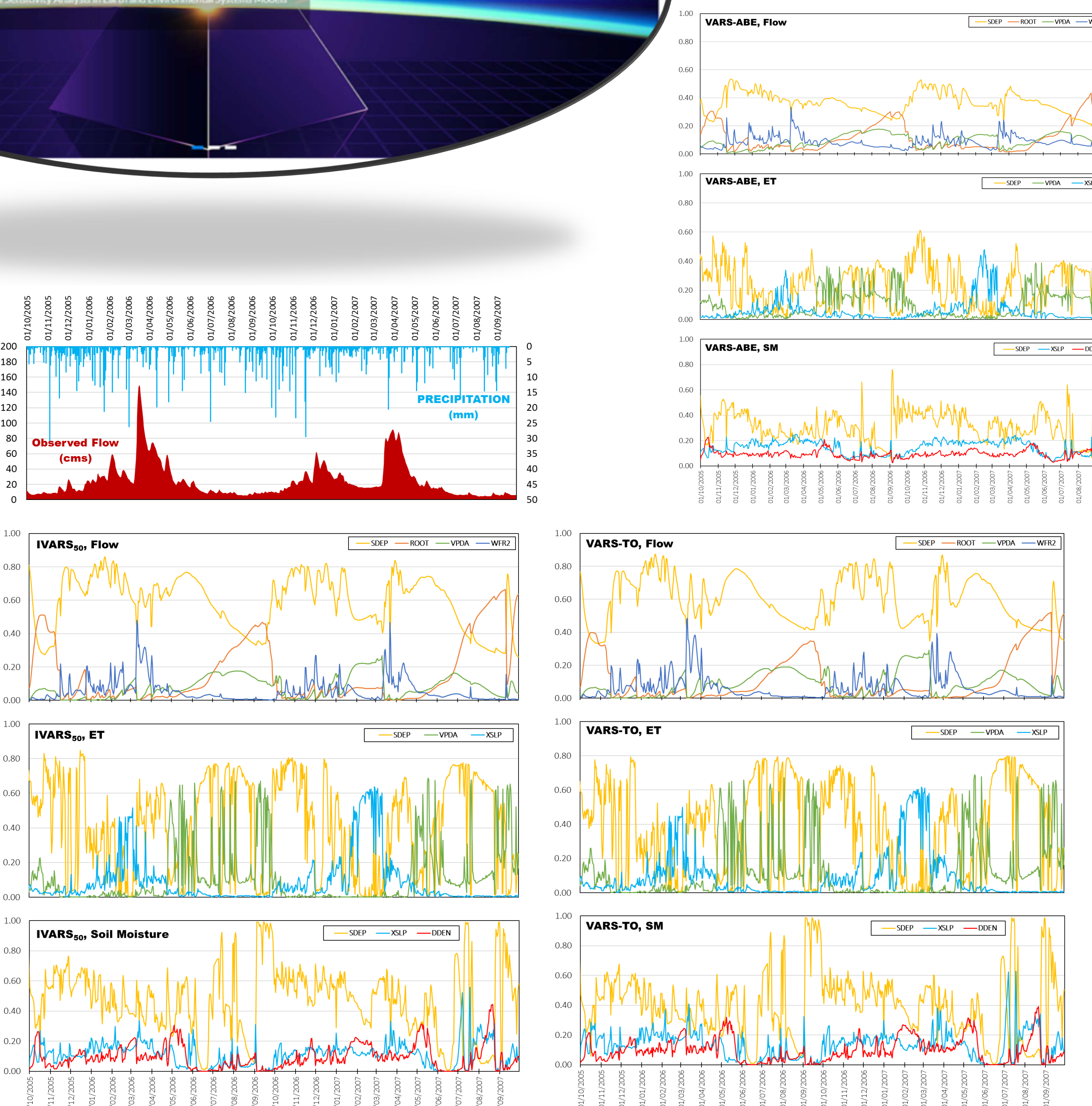
VARS-TOOL (vars-tool.com)



Time-Aggregate Sensitivity Indices



Time-Varying Sensitivity Indices



OBJECTIVES

To illustrate how STAR-VARS algorithm coupled with the GGSM approach facilitates a computationally-efficient comprehensive GSA using different methods, and how it enables learning about the temporal variability of dominant factors in response of distributed watershed models. For this purpose, we use the VARS-TOOL software toolbox (vars-tool.com), a comprehensive GSA toolbox, developed based on VARS (variogram analysis of response surfaces) approach.

CASE STUDY

Application of MESH (Modélisation Environnementale-Surface et Hydrologie) to Nottawasaga river basin in Canada. MESH is a semi-distributed physically-based coupled land surface-hydrology modelling system developed by Environment and Climate Change Canada (ECCC) for various water resources management purposes in Canada. MESH couples the Canadian land surface scheme (CLASS) with a routing module, WATROUTE.

CONCLUSIONS

- Based on GGSM approach, VARS-TOOL can efficiently produce series of sensitivity metrics based on multiple GSA methods.
 - This includes IVARS (variogram-based), VARS-TO (variance-based), and (VARS-ABE (derivative-based)).
 - Both time-varying and time-aggregate sensitivity metrics can be generated.
- All 3 GSA methods show similar sensitivity results for all three responses Flow, ET, and Soil moisture.
 - SDEP, ROOT, VPDA, and XSLP are the most influential parameters.
 - Model sensitivity to parameters vary significantly with time. SDEP and XSLP tend to be more influential during higher flows. ROOT and VPDA become more important during the crop growing season and lower flows.

REFERENCES

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- Razavi S., Sheikholeslami R., Gupta H., Haghnegahdar A., (2019), VARS-TOOL: A toolbox for comprehensive, efficient, and robust sensitivity and uncertainty analysis, *Environmental Modelling & Software*, Volume 112, 95-107, <https://doi.org/10.1016/j.envsoft.2018.10.005>.
- Razavi, S., and Gupta, H. V., (2016), A new framework for comprehensive, robust, and efficient global sensitivity analysis: II. Application, *Water Resources Research*, 51, doi:10.1002/2015WR017559.

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