Invitation

Public PhD defense of Robin Thibaut

Location

Ghent, Campus Sterre
Building S9
Auditorium A2

Monday 27 March 2023

16:00 - 18:00

Streaming link

A reception is held after the presentation.

To confirm your attendance at the reception, send a message to robin.thibaut@ugent.be
by March 14.

Machine Learning for Bayesian Experimental Design in the Subsurface

Dissertation submitted in fulfilment of the requirements for the Degree of

Doctor of Science: Geology

Ghent University



Advisors

Prof. Dr. Thomas Hermans
Ghent University

Dr. Eric Laloy Belgian Nuclear Research Centre

Jury members

Prof. Dr. Ellen Van de Vijver Ghent University

Prof. Dr. Frédéric Nguyen University of Liège

Prof. Dr. Kristine Walraevens
Ghent University

Prof. Dr. Ty Ferré University of Arizona

Chairman

Prof. Dr. Stephen Louwye Ghent University





Personalia



Robin Thibaut is a PhD fellow in Applied Geology and Hydrogeology at Ghent University, Belgium. His research focuses on developing an experimental design framework for Earth Sciences, leveraging his exper-

tise in machine learning and open-source software development.

Since 2014, Robin has earned one bachelor's and one master's degree in Geological Sciences and Geological Engineering at the Université Libre de Bruxelles and the University of Liège, respectively, and has published three A1 peer-reviewed papers as first author.

He also worked as a Project Engineer at an offshore survey company and participated in a geophysical mission to Cambodia to assess gold deposits. Passionate about travelling, Robin has a working knowledge of English, Vietnamese, Dutch and French, and is proficient in multiple programming languages such as Python.

coding!





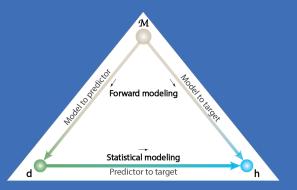




Research Highlights

Presentation of a novel framework for Bayesian optimal experimental design that integrates simulation and data-driven methods, Bayesian inference, and machine learning.

The framework is based on Bayesian Evidential **Learning**, a Monte Carlo method that employs machine learning to learn a direct relationship between predictor and target variables.



The framework is adaptable and can accommodate a wide range of data sources, including field observations and simulated data.

Three case studies in groundwater modeling have demonstrated the efficacy of the framework:

- wellhead protection area delineation
- aquifer thermal energy storage monitoring
- groundwater-surface water interaction

Publication of several open-source Python packages and datasets.