# Joint data assimilation and parameter calibration in real-time groundwater modelling using nested particle filters

Maximilian Ramgraber<sup>1,2</sup>, Mario Schirmer<sup>1,2</sup>

- <sup>1</sup> Department Water Resources & Drinking Water, Swiss Federal Institute of Aquatic Science and Technology (Eawag), Switzerland
- <sup>2</sup> Centre d'hydrogéologie et de géothermie (CHYM), University of Neuchâtel, Switzerland

## Introduction

## Motivation:

groundwater modelling is notoriously limited by availability of geological data

· inverse parameter estimation with batch of available data

computational effort; with new data. lengthy recalibration is required

### **Alternative**

sequential (real-time) calibration

## State of the Art: EnKF

ensemble Kalman filter (EnKF) solves Bayesian inference in a special case. For calibration:

## Required assumptions:

- · Gaussian distributions
- · linear model dynamic
- · states and parameters are jointly Gaussian

Hydrogeology: assumptions not met → stochastic significance lost

Calibration is Gaussian

→ too simple to preserve geology

- Joint Gaussianity
- → parameters **implictly** time-varying

Assumption of time-varying parameters does not entail maior errors in dissipative settings, which tend to 'forget' their history.

# Proposed approach

Employ particle filters, a general solution to Bayesian inference requiring fewer assumptions than the EnKF:

Curse of dimensionality → stochastic significance questionable



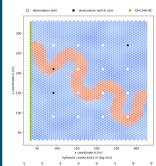
Artificial parameter dynamic (kernel) → parameters explicitly time-varying

## How do nested particle filters work? The inner particle filters The nested structure separates inference for provide likelihoods for Each particle the outer particle filter. states and parameters. assimmilates data individual performance varies Inner particle filter: Data assimilation for model states Surviving Poor particles particles are are discarded. mutated, creating good particles are Outer particle filter: new diversity duplicated Auto-calibration of model parameters This is where the parameter dynamic comes in.

# Model setup

# Synthetic reality:

- 2-D aquifer
- 1 m fixed head western boundary
- sinusoidal recharge
- 16 observation wells (obs. error 2 cm)
- · 2850 hexagonal cells
- implemented in MODFLOW USG
- geology: highly conductive meander
- facies type known at three wells
- boundary conditions assumed known



# Artificial parameter dynamic

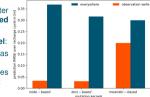
- hyper(parameter)space:
- describe desired geology via hyperparameters
- hyperspace can be projected into parameter space
- applying dynamic in hyperspace preserves geological patterns
- curse of dimensionality alleviated in lower-dimensional hyperspace

# Hyperspace: number of corners +2 corners Realspace: desired dynamic

## Discussion

## Discussion:

- promising performance of parameter estimation for node- and lens-based
- high RMSE of meander-based kernel:
  - · Facies map anisotropy was deliberately misspecified
  - lens-based kernel compensates via off-meander lens placement



## Augmented reality

Watch the algorithm in action

## Outlook:

- · model self-diagnosis: Investigate parameter surrogacy to identify structural errors
- · complex hyperspace kernels: Investigate possibility for calibrating more detailed geological structures

# Results from different hyperparameter kernels



~50 nodes

Hyperparameters:

· number of nodes

· node positions and values



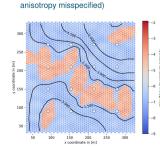
· projection via inverse distance weighting



- ~12 elliptic lenses
- · projection via structure mapping

## Hyperparameters:

- · number of lenses
- · lens geometries (rotation, size, aspect)
- · facies maps (hydraulic conductivities.



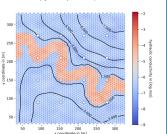


 one meander projection via structure mapping

## Hyperparameters:

## · meander geometry (number of turns.

- start & end, meander & channel width)
- facies maps (hydraulic conductivities. anisotropy misspecified)





AR instructions

Scan any of the QR

codes in section (E).

allow camera access. focus this marker, tap screen to play video.