

# **The application of Bayesian approaches in water quality modelling**

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**HS2.3.3/BG4.28  
EGU General Assembly 2020**

# Presentations



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Key challenge	Key finding
<b>Craig Stow: Process based or probabilistic models?</b> <a href="https://doi.org/10.5194/egusphere-egu2020-9925">https://doi.org/10.5194/egusphere-egu2020-9925</a>	
<ul style="list-style-type: none"><li>• Are complex models better?</li><li>• The need for explicit uncertainty analysis of process-based models</li><li>• Specification of prior distribution</li><li>• Computational challenges</li></ul>	<ul style="list-style-type: none"><li>• The benefits of Bayesian approaches</li></ul>
<b>Song Qian: A normative definition of a Bayesian prior</b> <a href="https://doi.org/10.5194/egusphere-egu2020-17978">https://doi.org/10.5194/egusphere-egu2020-17978</a>	
<ul style="list-style-type: none"><li>• How to derive and formulate a prior distribution?</li><li>• Prescriptive definition of a Bayesian prior</li></ul>	<ul style="list-style-type: none"><li>• Two case studies presented:<ul style="list-style-type: none"><li>• Modelling of cyanobacterial toxins</li><li>• Improvement of chemical calibration curve</li></ul></li></ul>

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Key challenge	Key finding
<b>George Arhonditsis: Castles built on sand or predictive limnology in action?</b> <b>The importance of Bayesian ensembles to support our ecological forecasts</b> <a href="https://doi.org/10.5194/egusphere-egu2020-7836">https://doi.org/10.5194/egusphere-egu2020-7836</a>	
<ul style="list-style-type: none"><li>• Many different model structures and many different parameter sets within a chosen model structure can acceptably reproduce the observed behavior of a complex environmental system</li></ul>	<ul style="list-style-type: none"><li>• Need to adopt a multi-model strategy rather than the single “best-fit” model Present a methodological framework to develop multi-model ensembles</li><li>• Implemented framework on 2 cases studies</li></ul>
<b>Yong Liu and Sifeng Wu: Resilience indicator for ecosystems subject to high risk of irreversible degradation: a probabilistic method based on Bayesian inference</b> <a href="https://doi.org/10.5194/egusphere-egu2020-6182">https://doi.org/10.5194/egusphere-egu2020-6182</a>	
<ul style="list-style-type: none"><li>• Ecosystem degradation is usually abrupt with unexpected shifts</li><li>• Some ecosystems might be subject to high risks of irreversible degradation because of strong undesirable resilience</li></ul>	<ul style="list-style-type: none"><li>• A practical framework to identify sensitive regions for conservation as well as opportunities for mitigation</li><li>• Method implemented on lake eutrophication</li></ul>

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<b>Daniel Obenour et al.:</b> <b>Assessing within-lake nutrient cycling through multi-decadal Bayesian mechanistic modeling</b> <a href="https://doi.org/10.5194/egusphere-egu2020-4232">https://doi.org/10.5194/egusphere-egu2020-4232</a>	
<ul style="list-style-type: none"><li>• <b>Bayesian calibration of a mechanistic model to understand nutrient recycling from lake bottom sediments</b></li><li>• <b>Combine mass-balance model with Bayesian inference</b></li></ul>	Nutrients stored in lacustrine sediment are an important source of internal loading to the reservoir for multiple decades, and will dampen the effects of external watershed loading reductions
<b>Ibrahim Alameddine and Eliza Deutsch:</b> <b>Understanding Harmful Algal Bloom Dynamics in a Mediterranean Hypereutrophic Reservoir insights from a Bayesian Network and a Structural Equation Model</b> <a href="https://doi.org/10.5194/egusphere-egu2020-6709">https://doi.org/10.5194/egusphere-egu2020-6709</a>	
<ul style="list-style-type: none"><li>• <b>Identifying pathways between the physical lake conditions and the nutrient loads on one hand and ecological endpoint on the other</b></li><li>• <b>Comparing BN and SEM model structures</b></li></ul>	<p>Prior model structure not supported by data</p> <p>Models largely concur in structure</p> <p>Both models capture temperature effects and direct nutrient pathways and highlight the importance of internal loading</p>

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<b>Danlu Guo et al.:</b> <b>A Bayesian hierarchical model to predict spatio-temporal variability in river water quality at 102 catchments</b> <a href="https://doi.org/10.5194/egusphere-egu2020-4725">https://doi.org/10.5194/egusphere-egu2020-4725</a>	
<b>Challenges to explain temporal variability in water quality using statistical models</b> <b>Linear statistical models are limited in representing water quality datasets with large proportions of below-detection-limit records</b>	Model improvements should focus on: <ul style="list-style-type: none"><li>• Alternative statistical model structures to improve fitting for truncated data</li><li>• Better representation of non-conservative constituents by accounting for biogeochemical processes</li></ul>
<b>Minkyu Jung et al.:</b> <b>A Hierarchical Bayesian Model for Spatio-Temporal Water Quality Modeling in a Changing Climate in South Korea</b> <a href="https://doi.org/10.5194/egusphere-egu2020-21271">https://doi.org/10.5194/egusphere-egu2020-21271</a>	
<b>Difficult to obtain accurate predictions of water quality due to the large spatio-temporal variability in a changing climate</b>	<b>Hierarchical Bayesian model</b> can capture the key aspects of the water quality parameters in terms of seasonality and their uncertainty

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<b>Lorenz Ammann et al.:</b> <b>Patterns in time-dependent parameters reveal deficits of a catchment-scale herbicide transport model</b> <a href="https://doi.org/10.5194/egusphere-egu2020-9081">https://doi.org/10.5194/egusphere-egu2020-9081</a>	
<b>Deterministic dynamic water quality models are too rigid: they do not allow for the stochastic nature of the system and are susceptible to structural errors</b>	Introducing stochasticity through time-dependent parameters can reveal deficits in model structure and can allow for a better description of the intrinsic uncertainty of dynamic water quality models
<b>Sakari Kuikka:</b> <b>Experiences in applying Bayesian network models in interdisciplinary water quality decision analysis</b> <a href="https://doi.org/10.5194/egusphere-egu2020-7270">https://doi.org/10.5194/egusphere-egu2020-7270</a>	
<b>Developing integrative Bayesian models in interdisciplinary analysis</b> <b>Different traditions and quality criteria of different scientific fields create both technical and human challenges to the modelling tasks</b>	Applications are based mainly on the use of expert knowledge, especially for decision options that have not been applied before <b>Bayesian decision analysis for management provides scientifically justified uncertainty estimates</b>

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<b>Camilla Negri et al.:</b> <b>Modelling phosphorus pollution risk in agricultural catchments using a spatially distributed Bayesian Belief Network</b> <a href="https://doi.org/10.5194/egusphere-egu2020-555">https://doi.org/10.5194/egusphere-egu2020-555</a>	
<b>Diffuse pollution of phosphorus (P) from agriculture is a major pressure on water quality</b> <b>Need to develop Decision Support Tools that can account for the uncertainty in both data and models</b>	Model captures the difference in P loss risk between catchments, probably caused by contrasting hydrological characteristics and soil P sources. Climate change and land use change scenarios crucial to inform targeting of mitigation measures
<b>Magnus Norling: Rapid development and evaluation of fast process-based models in Mobius</b> <a href="https://doi.org/10.5194/egusphere-egu2020-7326">https://doi.org/10.5194/egusphere-egu2020-7326</a>	
<b>Build and explore many model structures and evaluate model uncertainty</b>	<b>Modelling frameworks are a good alternative to one-size-fits-all models</b> , and we hope Mobius will be a useful tool for promoting more robust modelling



# Discussion



# Discussion Points

Challenge	Question	Examples
<b>Model complexity and uncertainty assessment</b>	<b>Do we need simpler models, faster models, or both?</b> Developments in computational capacity have led to more complex models, not necessarily to better predictive performance	Craig Stow Magnus Norling
<b>Prior distributions</b>	<b>What is a good prior?</b> Expert elicitation, informative, and non-informative priors	Song Qian Daniel Obenour
<b>Model structural uncertainty</b>	<b>Is our model structure adequate?</b> Model ensembles, flexible and fast frameworks for controlled model comparison, flexibility in model structure through time-dependent parameters	George Arhonditsis Ibrahim Alameddine Magnus Norling Lorenz Ammann
<b>Representing spatio-temporal variability in models</b>	<b>Is our input data adequate?</b> <b>How do you decide on your spatio-temporal scale?</b> Data resolution – spatial & temporal, uncertainty in model predictions	Minkyu Jung Danlu Guo

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<b>Need for decision support tools with explicit uncertainty quantification</b>	<b>What is the way forward in using models for decision support?</b> <b>Are we effectively integrating uncertainties in our decision making process?</b> <ul style="list-style-type: none"><li>• Different traditions and quality criteria from different scientific fields (biology, sociology and environmental economics) create both technical and human challenges to the modelling tasks</li><li>• Bayesian decision analysis for management provides scientifically justified uncertainty estimates</li></ul>	Sakari Kuikka Camilla Negri Craig Stow
<b>Ecological system complexity and resilience – impacting effectiveness of mitigation interventions</b>	<b>How best to simulate complex biophysical systems?</b> <ul style="list-style-type: none"><li>• Accounting for unexpected shifts in ecosystem states</li><li>• Modelling nutrient recycling from sediments</li><li>• Identifying pathways and feedbacks between drivers and response variables</li></ul>	Yong Liu Daniel Obenour Ibrahim Alameddine

# Important announcements

**ANNOUNCEMENT** 

- Thank you for supporting this session in this **EXCEPTIONAL** year – we look forward to meeting you in person in Vienna at EGU 2021!

- **SPECIAL ISSUE ALERT**

- We are proposing a *Special Issue* on '**Frontiers in the application of Bayesian approaches in water quality modelling**' in the EGU Hydrology and Earth Systems Science Journal
- Open both to presenters at this session over the past two years and to the wider community
- Interested to contribute to the Special Issue?  
Please get in touch with [miriam.glendell@hutton.ac.uk](mailto:miriam.glendell@hutton.ac.uk)