

Effects of Inhomogeneity and (Statistical and Material) Anisotropy on THM simulations

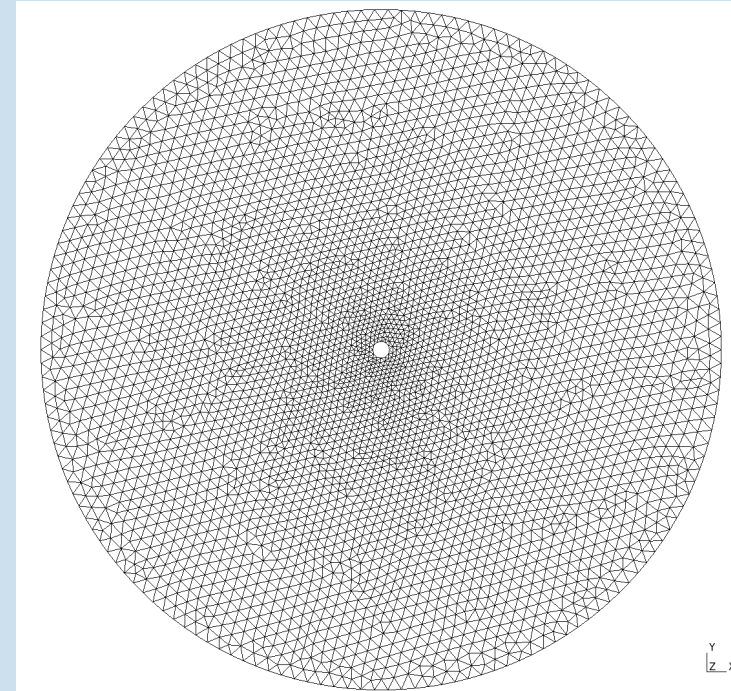
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Motivation and highlights

- Investigations into the integrity of barriers [Kurgis et al. 2024]
- Coupled numerical THM simulations with OpenGeoSys-6 [Bilke et al. 2022]
- Consideration of the uncertainty of input parameters
- Effects of the transverse isotropy of the input parameters
- Input parameters as inhomogeneous random fields
- Consideration of statistical and material anisotropy

Model setup and specifics

- Simplified 2D mesh based on the FE experiment [Müller et al. 2018]
- Host rock (Opalinus Clay) → $d = 100$ m
- Circular heat source with a diameter of 2.48 m → emplaced waste cell
- Thermal conductivity (λ), intrinsic permeability (k), and elasticity moduli (E_i) as uncertain input parameters
- Generation of inhomogeneous random fields with `kleme`
- Effects on temperature, pore water pressure, displacement, Darcy velocity, von-Mises, and hydrostatic effective stresses.



Initial and boundary conditions:
 $T_0 = 15^\circ\text{C}$, $p_0 = 2$ MPa, $u_{S0} = 0$
 $\vec{q} \cdot \vec{n}$ at the tunnel boundary
 $p = 0$ at the tunnel boundary
 $\text{Normal } \vec{u}_S = \vec{0}$ at the outer boundary

Generation of random fields through the `kleme` library

- The permeability field is assumed to be log-Gaussian;
- Realizations of Gaussian random fields are generated by the truncated Karhunen-Loève expansion, which approximates a random field as

$$\mathbf{Z}(\mathbf{x}, \xi) \approx \bar{\mathbf{Z}}(\mathbf{x}) + \sum_{i=1}^M \xi_i \sqrt{\lambda_i} \mathbf{f}_i(\mathbf{x})$$

where $\bar{\mathbf{Z}}(\mathbf{x})$ is the mean of the field, $\{\xi_i\} \sim \mathcal{N}(0, 1)$, and $\{\mathbf{f}_i\}$ and $\{\lambda_i\}$ are eigenfunctions and eigenvalues of the covariance operator

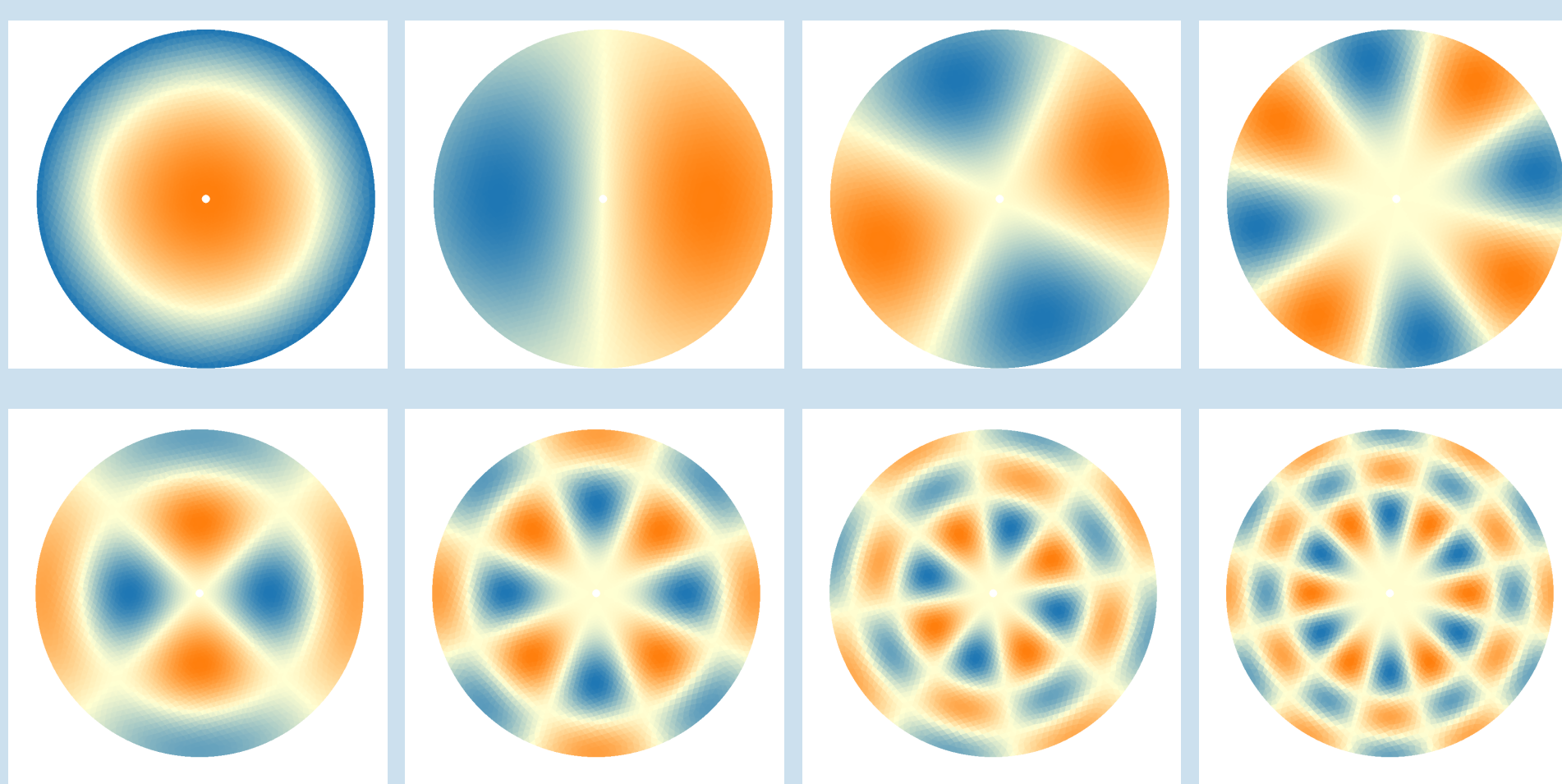
$$(Cu)(\mathbf{x}) = \int_D c(\mathbf{x}, \mathbf{y}) u(\mathbf{y}) d\mathbf{y}$$

and $c(\mathbf{x}, \mathbf{y})$ is the covariance function;

- The Galerkin method is used to discretize the covariance operator and then solve for the eigenfunctions and eigenvalues;
- The exponential covariance function is used in this study

$$c(\mathbf{x}, \mathbf{y}) = \sigma^2 \exp\left(-\frac{|\mathbf{x} - \mathbf{y}|}{l}\right)$$

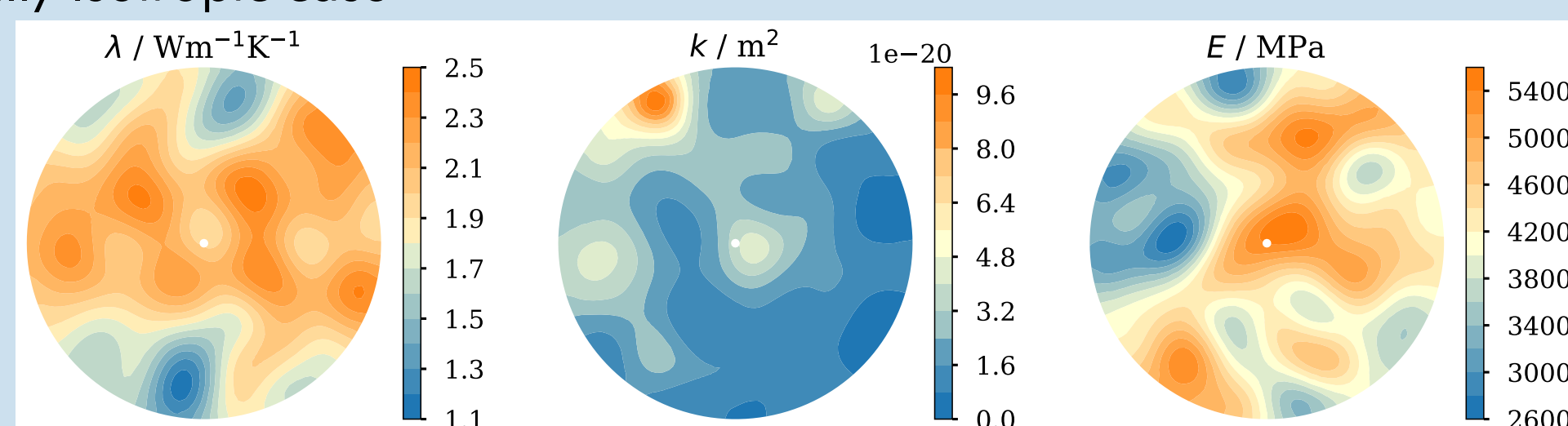
where σ is the scale factor and l is the correlation length.



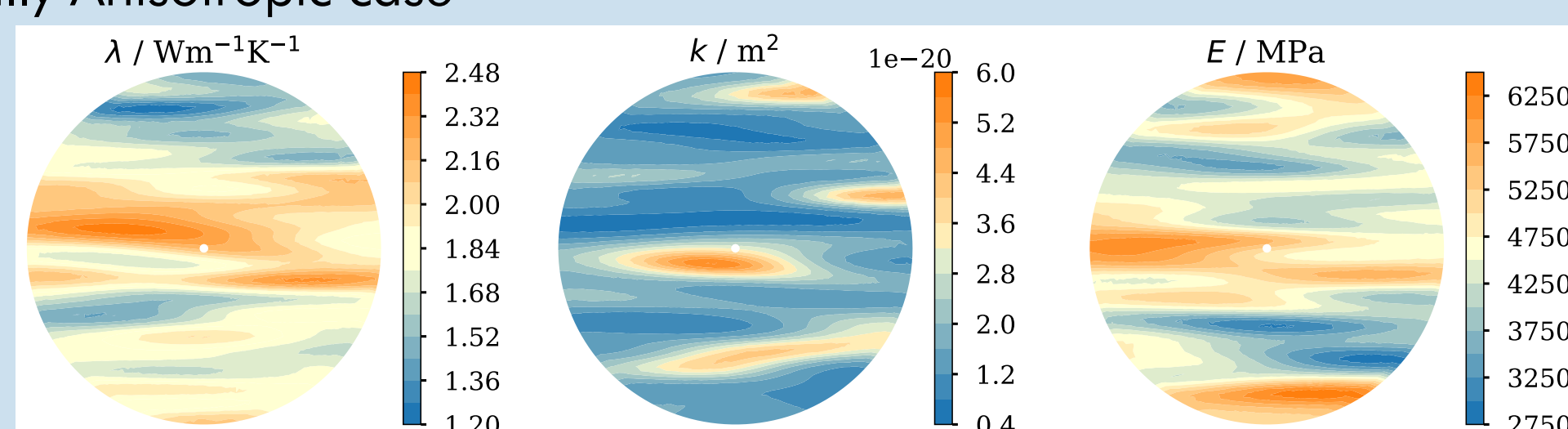
Eigenfunctions 1, 2, 5, 11, 13, 25, 43, and 64 of the domain of interest, using a Matérn covariance function with smoothness parameter $\nu = 0.5$ and characteristic length $\ell = 15$ m. The outer diameter of the domain is set to 100 m. Orange hues indicate higher values and a colorbar is omitted as the relative magnitudes of the eigenfunctions are not of primary interest in this context.

Examples of inhomogeneous random fields

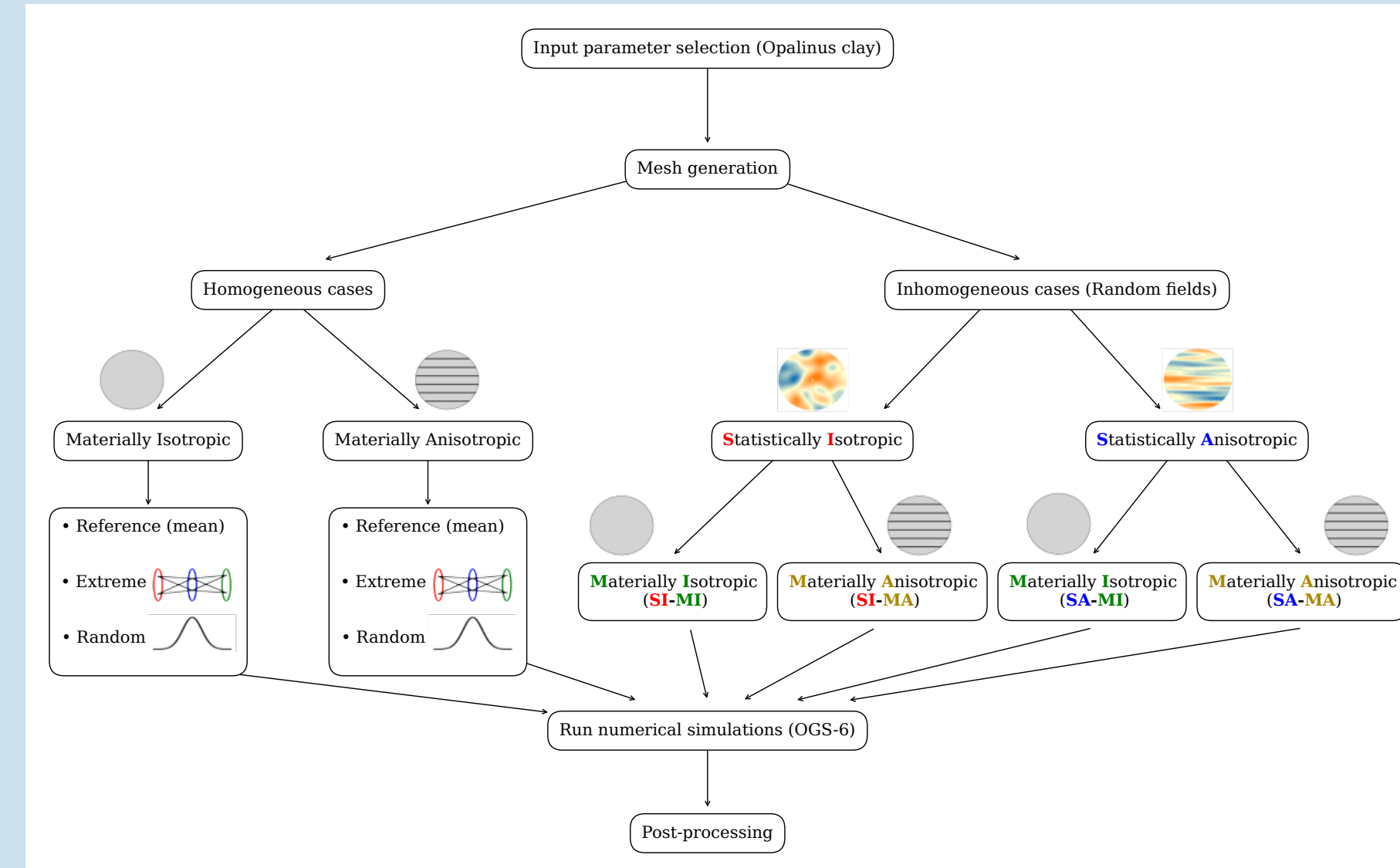
Statistically Isotropic case



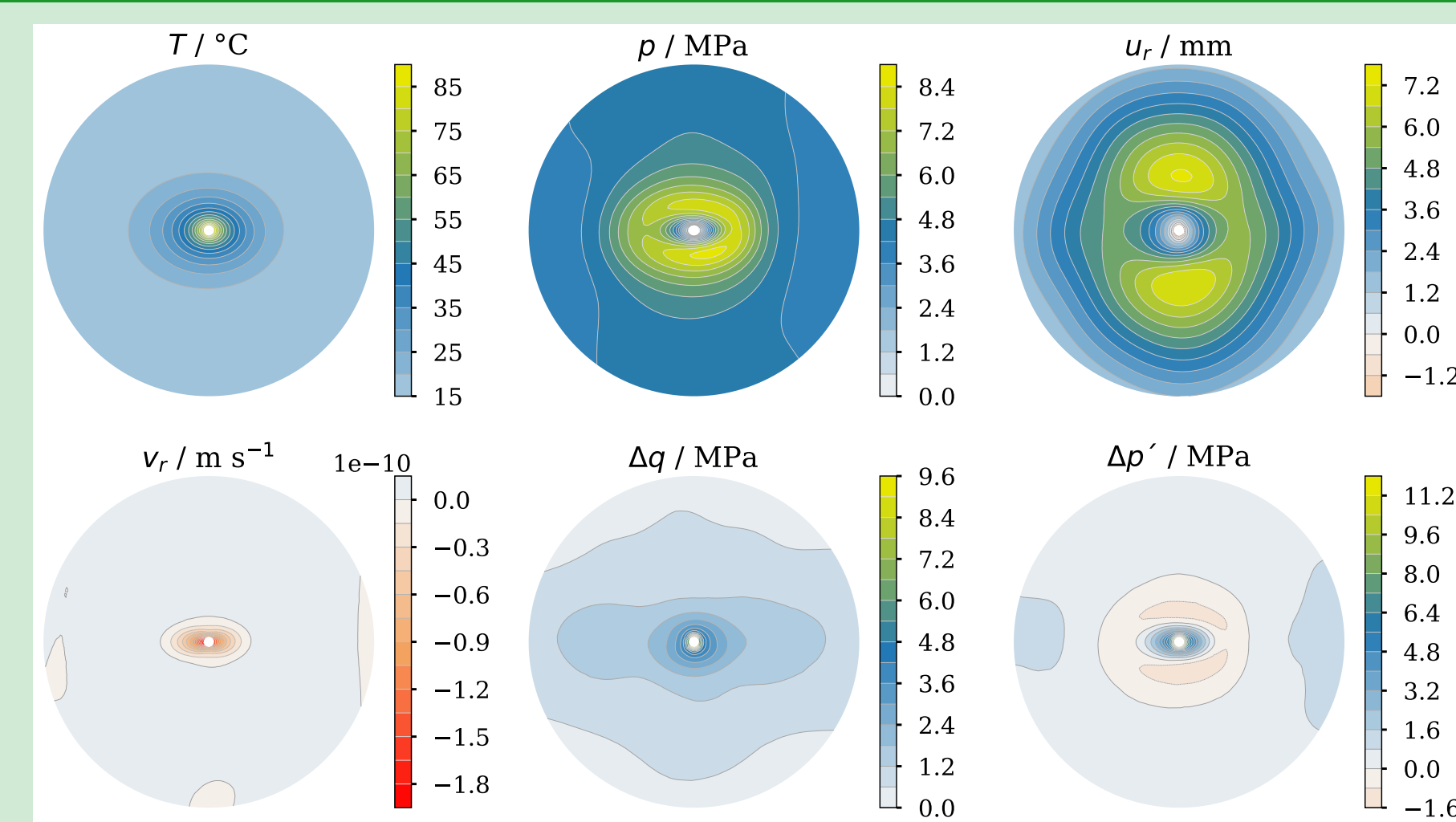
Statistically Anisotropic case



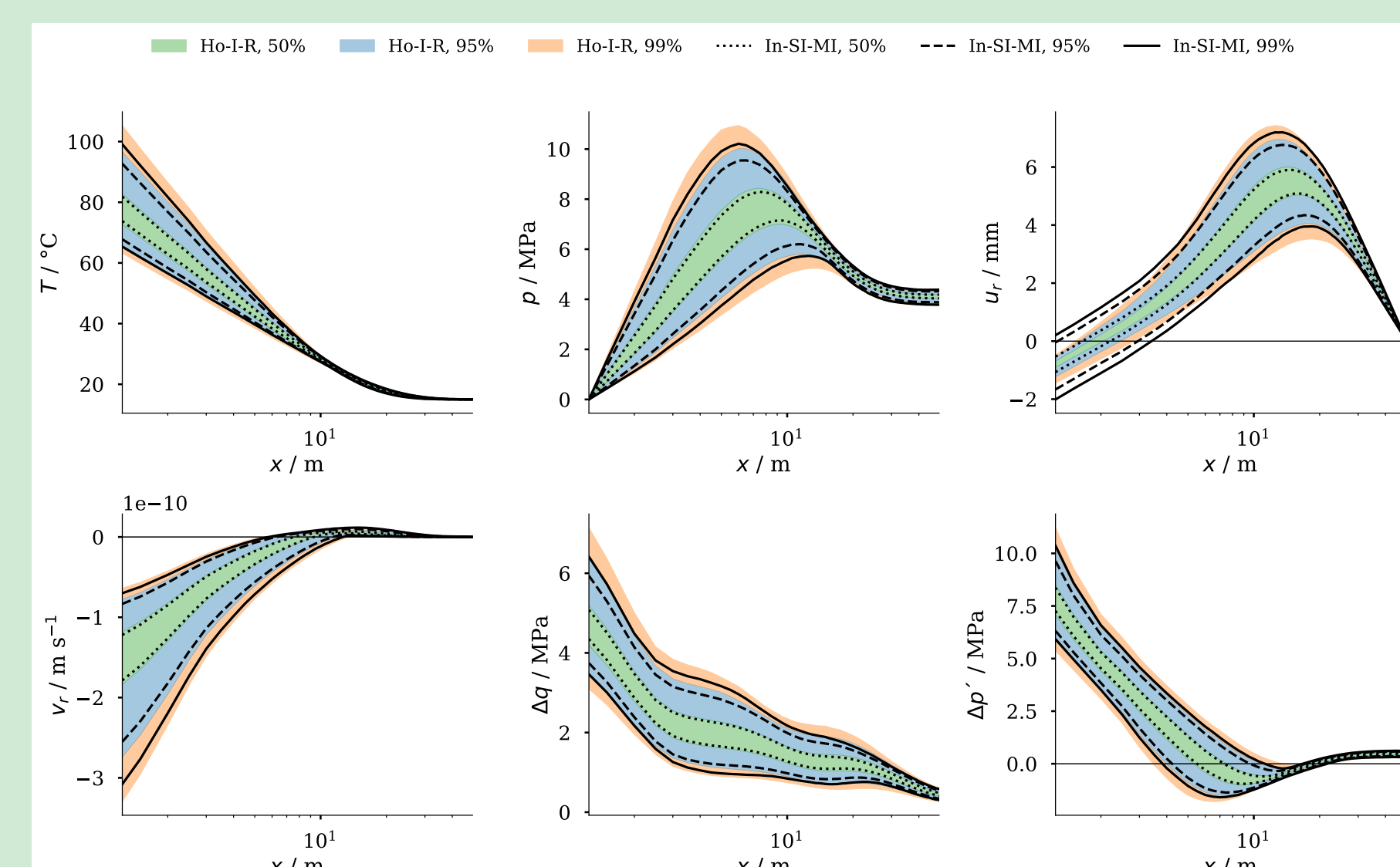
Workflow



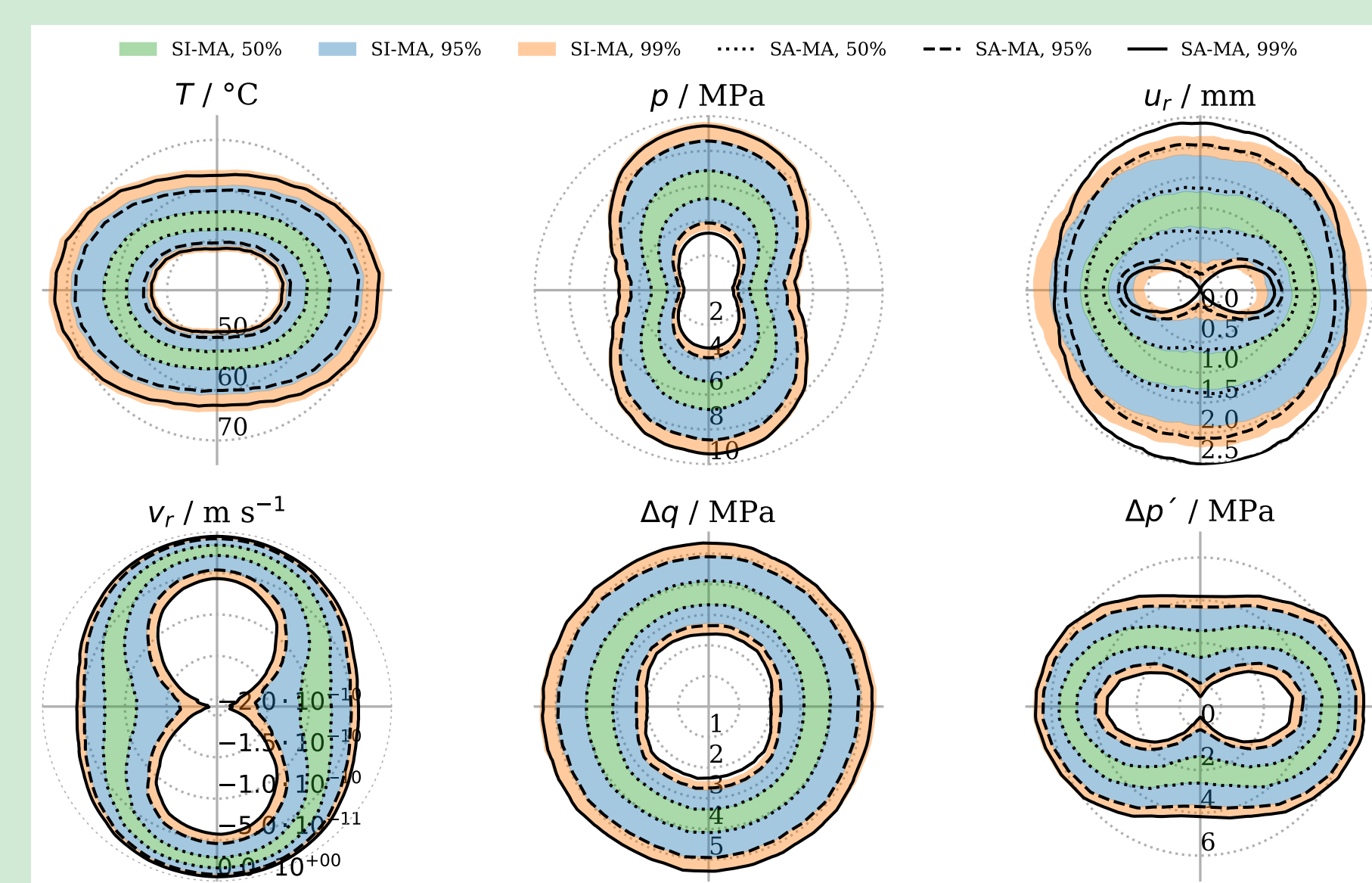
Results



Contour plot for statistically isotropic and materially anisotropic case using one random realization of a random field for input parameters up to a radius of $r = 40$ m



Plot showing comparison of various inter-percentile ranges (IPR) of output variables for 10000 realizations of random fields of input parameters along the x -axis for the homogeneous-isotropic (Random) and inhomogeneous-isotropic (SI-MI) case.



Polar plot showing 50%, 95% and 99% IPR for the statistically isotropic and anisotropic cases in the presence of material anisotropy for all output variables at a distance of 3 m from the heat source.

Conclusions

- The temperature development is not influenced by inhomogeneity or anisotropy of the intrinsic permeability and stiffness; see also Chaudhry et al. 2021.
- The effect of statistical anisotropy is visible even in the absence of material anisotropy.
- Stress-based indicators show that material anisotropy has a more pronounced effect than statistical anisotropy.
- Interpretation of results with regard to integrity criteria.

Outlook

- Introduce varying statistical anisotropy ratios (instead of fully correlated fields).
- Include additional uncertain parameters (e.g., bentonite layers).
- Validate and calibrate models using field-scale data.
- Investigate surrogate modeling to reduce computational costs.

References:

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- Aqeel Afzal Chaudhry, Jörg Buchwald, and Thomas Nagel. "Local and global spatio-temporal sensitivity analysis of thermal consolidation around a point heat source". In: *International Journal of Rock Mechanics and Mining Sciences* 139 (2021), p. 104662. ISSN: 1365-1609.
- Kata Kurgis, Peter Ahtziger-Zupančič, Merle Bjorge, Marc S Boxberg, Matteo Broggi, Jörg Buchwald, Oliver G Ernst, Judith Flügge, Andrey Ganopolski, Thomas Graf, et al. "Uncertainties and robustness with regard to the safety of a repository for high-level radioactive waste: introduction of a research initiative". In: *Environmental Earth Sciences* 83.2 (2024), p. 82.
- Herwig R Müller, Benoît Garitte, Tobias Vogt, Sven Köhler, Toshihiro Sakaki, Hanspeter Weber, Marian Hertrich, Jens K Becker, Niels Giroud, et al. "Implementation of the full-scale emplacement (FE) experiment at the Mont Terri rock laboratory". In: *Mont Terri Rock Laboratory, 20 Years: Two Decades of Research and Experimentation on Claystones for Geological Disposal of Radioactive Waste* (2018), pp. 289–308.