Simulation-Based Inference for Geothermal Models







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Background

- Geothermal model calibration involves adjusting parameters, including rock permeabilities and hot mass upflows, until the model temperatures and pressures match historical data.
- Bayesian model calibration begins with the *prior* distribution, which encodes one's beliefs in the values of the parameters prior to seeing data, then data is conditioned on to give the posterior distribution, which characterises the remaining uncertainty in the parameters.
- Accurately characterising the posterior of a geothermal model allows for better quality decisions to be made in terms of the management of the real geothermal system. Generally, linear approximations are used; however, simulation-based methods are potentially more accurate.

Methods

Two simulation-based methods were used to characterise the posterior of a synthetic geothermal model.

ABC (approximate Bayesian computation) Description: models with parameters sampled from the prior are run, their fit to the data is evaluated, and parameters of the best-fitting models are used to characterise the posterior.

▶ 1000 models were run, and the best-fitting 5% were used to characterise the posterior.

Population ABC

Description: same as ABC, but populations of models are used; the first is sampled from the prior, but the rest are formed by modifying the best models in the previous population. This can be more efficient than ABC.

• After the first population, most models failed to converge. A decision tree was fit to determine why.

Conclusions

ABC

- Even the best-fitting models do not fit the data well (figure 1).
- Consequently, the ABC posterior is still similar to the prior, but very different to a standard linear approximation (figure 2).
- Unless a large number of models are run, which is unrealistic for most geothermal models, ABC will characterise the posterior poorly.

Population ABC

- The decision tree suggests a potential cause of the models failing is some of the permeabilities being too high (figure 3).
- Bounding the range of acceptable permeabilities reduces the failure rate significantly.
- The algorithm is currently being run with these additional constraints and will finish in the coming weeks.

3 Results

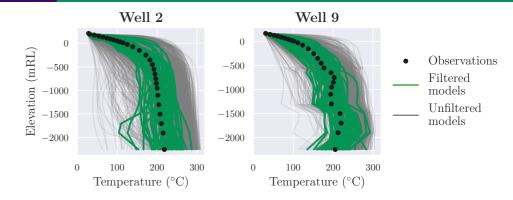


Figure 1. ABC model temperature predictions for selected wells.

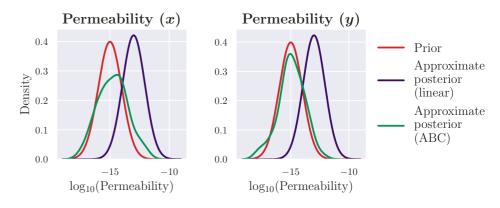


Figure 2. Prior (red), and linear / ABC posterior distributions (purple / green), for permeabilities of the IAC01 rock type.

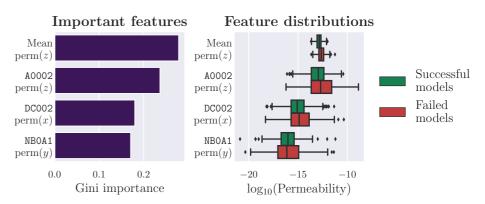


Figure 3. Most important features of decision tree (*left*), and the distributions of these for successful / failed models (right).