

Peer-reviewed Conference Contribution

# Interpretation of Thermal Response Test (TRT) in energy piles using Bayesian Inference

Norma Patricia López-Acosta<sup>1,\*</sup> and David Francisco Barba-Galdámez<sup>1</sup>

<sup>1</sup>Instituto de Ingeniería, UNAM

\* Corresponding author: [NLopezA@iingen.unam.mx](mailto:NLopezA@iingen.unam.mx)

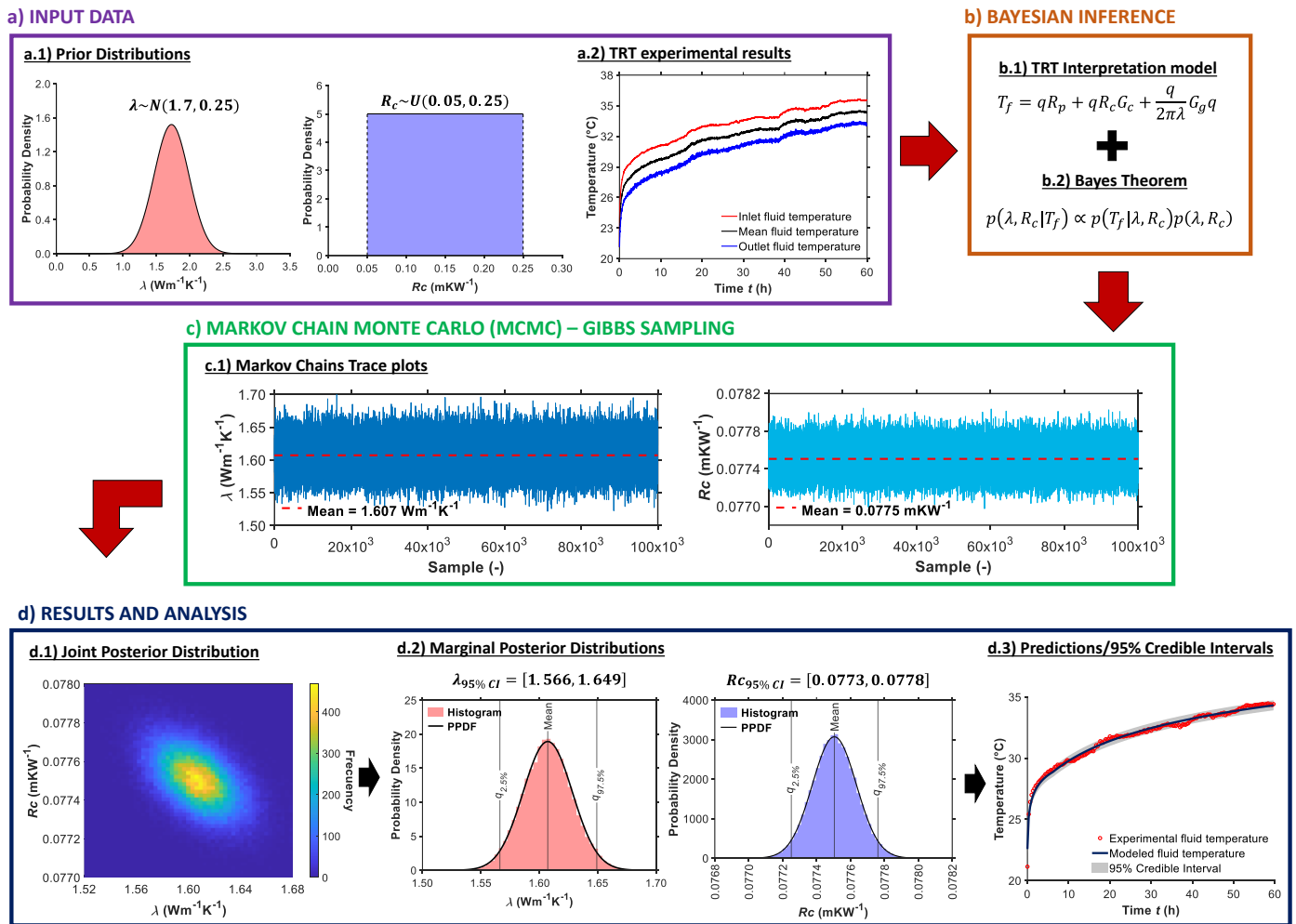
An efficient design of energy geo-structure systems requires an accurate characterization of the structure and ground thermal parameters (e.g., thermal resistance of the ground heat exchanger  $R_b$ , and the thermal conductivity of the surrounding ground  $\lambda$ ). The Thermal Response Test (TRT) is the international standard method for these parameter determinations. Traditionally, TRTs have been performed only in boreholes and piles of small diameter, i.e., 300 mm maximum. However, in recent years, TRTs in large-diameter piles are becoming more common. Although these types of TRTs can provide acceptable results if adequate interpretation methods are selected, some concerns about the added uncertainty of the estimations due to the increased logistic challenges prevail [4, 7].

In this context, Bayesian Inference represents an invaluable resource for analyzing TRTs in energy piles. Bayesian approaches allow accounting for structural model errors and provide credible intervals, giving uncertainty metrics that are difficult to obtain by classic deterministic methods. Additionally, since Bayesian techniques combine prior knowledge with observations (experimental data), they have a significant potential for uncertainty assessment in short-duration tests, as demonstrated by recent studies in traditional TRTs [1, 2, 8].

In the present study, a stochastic method based on Bayes' Theorem [1] has been adapted to interpret the results of a short-duration TRT carried out in an energy pile in Mexico [5]. The aim is to infer Posterior Probability Density Functions (PPDFs) for the parameters of interest (ground thermal conductivity  $\lambda$  and concrete thermal resistance  $R_c$ ), taking into account our previous knowledge (prior distributions) and the given measurements of the TRT (likelihood). For this case study, prior distributions were assigned based on the site stratigraphy and international ground thermal properties databases [3]. The [7] method was employed to calculate the likelihood. This method uses G-functions developed specifically for the ground ( $G_g$ ) and the pile ( $G_c$ ), which described their transient response. Finally, a Markov Chain Monte Carlo (MCMC) method (Gibbs sampling) was used to obtain the PPDFs. From the PPDFs, the point estimates of the parameters and their 95% credible intervals were extracted. The above procedure is described in detail in Figure 1. The results are consistent with those obtained using a deterministic analysis (see Table 1). However, the PPDF provides more information about the uncertainties of the parameters. The above is crucial to perform more realistic designs and to evaluate accurately the financial viability of a project based on energy geo-structures.

**Table 1: Results comparison between deterministic analysis and Bayesian Approach of a TRT executed in Mexico**

Parameter	Units	Deterministic analysis (López-Acosta <i>et al.</i> , 2022)	Bayesian Approach	
			Posterior Mean	95 % Credible Interval
Ground thermal conductivity $\lambda$	$\text{Wm}^{-1}\text{K}^{-1}$	1.630	1.607	1.566 – 1.649
Concrete thermal resistivity $R_c$	$\text{mKW}^{-1}$	0.078	0.0775	0.0773 – 0.0778



**Figure 2: TRT Interpretation using Bayesian Inference: a) Input data, b) Bayesian Inference, c) Markov Chain Monte Carlo – Gibbs sampling, and d) Results and analysis.** Note:  $\lambda$  = Ground thermal conductivity,  $R_c$  = Concrete thermal resistance,  $T_f$  = Fluid Temperature,  $q$  = Heat transfer rate,  $R_p$  = Pipe thermal resistance,  $G_g$  = Loveridge G-function for ground,  $G_c$  = Loveridge G-function for concrete, PPDF = Posterior Probability Density Function, CI = Credible Intervals,  $q_{2.5\%}$  = 2.5 % quantile,  $q_{97.5\%}$  = 97.5% quantile.

### Contributor statement

Conceptualization: Norma Patricia López-Acosta, David Francisco Barba-Galdámez; Formal Analysis: David Francisco Barba-Galdámez; Project Administration: Norma Patricia López-Acosta; Software: David Francisco Barba-Galdámez; Visualization: David Francisco Barba-Galdámez; Writing – Original Draft: David Francisco Barba-Galdámez; Writing- Review & Editing: Norma Patricia López-Acosta.

### References

- [1] Choi, W., Kikumoto, H., Choudhary, R., & Ooka, R. (2018). Bayesian inference for thermal response test parameter estimation and uncertainty assessment. *Applied Energy*, 209, 306–321
- [2] Choi, W., Menberg, K., Kikumoto, H., Heo, Y., Choudhary, R., & Ooka, R. (2018). Bayesian inference of structural error in inverse models of thermal response tests. *Applied Energy*, 228, 1473–1485
- [3] Dalla Santa, G., Galgaro, A., Sassi, R., Cultrera, M., Scotton, P., Mueller, J., Bertermann, D., Mendrinis, D., Pasquali, R., Perego, R., Pera, S., Di Sipio, E., Cassiani, G., De Carli, M., & Bernardi, A. (2020). An updated ground thermal properties database for GSHP applications. *Geothermics*, 85, 101758
- [4] Jensen-Page, L., Loveridge, F., & Narsilio, G. A. (2019). Thermal response testing of large diameter energy piles. *Energies*, 12(14), 1–25
- [5] López-Acosta, N.P., Martínez-Rivera, A.M., & Barba-Galdámez, D.F. (2022). First Thermal Response Test (TRT) for energy geo-structure applications in Mexico. In Rahman, M. & Jaksa, M. (Eds.) *Proceedings of the 20th International Conference on Soil Mechanics and Geotechnical Engineering* (pp. 4965 – 4970)
- [6] Loveridge, F., & Powrie, W. (2013). Temperature response functions (G-functions) for single pile heat exchangers. *Energy*, 57, 554–564
- [7] Loveridge, F., Powrie, W., & Nicholson, D. (2014). Comparison of two different models for pile thermal response test interpretation. *Acta Geotechnica*, 9(3), 367–384
- [8] Pasquier, P., & Marcotte, D. (2020). Robust identification of volumetric heat capacity and analysis of thermal response tests by Bayesian inference with correlated residuals. *Applied Energy*, 261, 114394