

Certified Dimension Reduction for Bayesian Updating with the Cross-Entropy Method

Max Ehre¹, Rafael Flock², Iason Papaioannou¹ and Daniel Straub¹

¹ Engineering Risk Analysis Group, Technical University of Munich, Theresienstrasse 90, 80333 Munich, Germany, max.ehre@tum.de, iason.papaioannou@tum.de, straub@tum.de.

² Department of Applied Mathematics and Computer Science, Technical University of Denmark, Asmussens Allé, 2800 Kgs. Lyngby, Denmark, raff@dtu.dk.

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Inverse problems are common in engineering: the parameters of a model often have to be estimated based on observations of the model response. The Bayesian approach has proven powerful for solving such problems; one formulates a prior distribution for the parameter state that is updated with the observations to compute the posterior parameter distribution. Solving for the posterior distribution can be challenging when, e.g., prior and posterior significantly differ from one another and/or the parameter space is high-dimensional. We use a sequence of importance sampling measures that arise by tempering the likelihood to approach inverse problems exhibiting a significant distance between prior and posterior. Each importance sampling measure is identified by cross-entropy minimization as proposed in the context of Bayesian inverse problems in [2]. To efficiently address problems with high-dimensional parameter spaces we set up the minimization procedure in a low-dimensional subspace of the original parameter space. The principal idea is to analyse the spectrum of the second-moment matrix of the gradient of the log-likelihood function to identify a suitable subspace. Following [1], an upper bound on the Kullback-Leibler-divergence between full-dimensional and subspace posterior is provided, which can be utilized to determine the effective dimension of the inverse problem corresponding to a prescribed approximation error bound. We suggest heuristic criteria for optimally selecting the number of model and model gradient evaluations in each iteration of the importance sampling sequence. We demonstrate the practical performance of this approach using examples from structural engineering set in various parameter space dimensions.

REFERENCES

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