

Normalizations of the Proposal Density in Markov Chain Monte Carlo Algorithms

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Abstract

We explore the effects of normalizing the proposal density in Markov Chain Monte Carlo algorithms, in the context of a nonlinear inverse problem. Our problem is that of reconstructing the conductivity term K in the 2-dimensional heat equation

$$u_{xx} + u_{yy} = \frac{2H}{K\delta}u \quad (1)$$

given temperatures at the boundary points, given by d . A Metropolis-Hastings MCMC algorithm is implemented to do so. Markov Chains produce a probability distribution of possible solutions conditional on the observed data. We generate a candidate solution K' and solve the forward problem, obtaining d' . In this way, at step n the probability of setting $K_{n+1} = K'$ is given by

$$\alpha(K'|K_n) \equiv \min \left\{ 1, \frac{P(K'|d)g(K_n|K')}{P(K_n|d)g(K'|K_n)} \right\} \quad (2)$$

For our given proposal density g , this is initially computed as

$$\alpha = \min \left\{ 1, e^{\frac{-1}{2\sigma^2} \sum_{i,j=1}^{n,m} [(d_{ij}-d'_{ij})^2 - (d_{ij}-d_{n_{ij}})^2]} \right\} = \min \{1, e^{-D}\} \quad (3)$$

We identify certain issues with this construction, stemming from large and fluctuating values of D . Using this framework, we develop normalization terms z_0, z and parameters λ that preserve the inherently sparse information at our disposal, rewriting (3) as

$$\alpha = z_0 e^{-\lambda z D} \quad (4)$$

We examine the results of this variant of the MCMC algorithm on the reconstructions of several 2-dimensional conductivity functions.

Key words: Ill-posed, Inverse Problems, MCMC, Normalization, Numerical Analysis.