

MAT 653: Statistical Simulation

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Simulated Annealing

Consider maximize $h(\theta)$ (global maximization):

Assume $h : \Theta \mapsto \mathbb{R}$, $\int_H h(\theta) < \infty$, $H(\theta) \propto \exp(\frac{h(\theta)}{T})$, $T > 0$, where T is the temperature.

Define the so-called Boltzman-Gibbs transformation of h to be

$$\pi_t(\theta) \propto \exp\left(\frac{h(\theta)}{T_t}\right).$$

As $T_t \downarrow 0$, the kernel of density will become more and more concentrated in a narrower neighborhood of the maximizer of h .

Algorithm

Suppose $g(\cdot)$ symmetric density function $g(-x) = g(x)$, e.g. $N(0,1)$.

1. simulate $\zeta^{(t)}$ from g
2. generate $\theta^{(t+1)}$ as

$$\theta^{(t+1)} = \begin{cases} \theta^{(t)} + \zeta^{(t)} & \text{with prob } \rho = \exp(\frac{\Delta h}{T_t}) \wedge 1, \Delta h = h(\theta^{(t)} + \zeta^{(t)}) - h(\theta^{(t)}) \\ \theta^{(t)} & \text{with prob } 1 - \rho \end{cases}$$

3. update $T_{t+1} \leftarrow T_t$. For example, one can let $T_t = \frac{1}{\log(1+t)}$ or $(\frac{1}{1.2})^t$.

Remarks:

1. If the perturbation increases h (i.e., $\Delta h \geq 0$), the new proposal is automatically accepted. If $\Delta h < 0$, then still accept it with certain probability $\rho(< 1)$.
2. When T_t is relatively large, many “bad” proposals (that is those $\Delta h < 0$) are still accepted. Therefore a larger part of the space can be explored. But at larger stages, as $T_t \downarrow 0$, that will limit the “bad” proposal that we allow.
3. When T_t is sufficiently small, one essentially “quenches” the process by accepting only “good” proposals (i.e., $\Delta h \geq 0$).