

MAT 653: Statistical Simulation

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Metropolis-Hasting algorithm (blockwise)

Target density: $f(x, y)$, which could be unnormalized.

We use two candidate transition densities $q_X(x|x_c, y_c)$ and $q_Y(y|x_c, y_c)$, where x_c, y_c stand for current values.

Algorithm:

1. Update X

(a) sample $X^* \sim q_X(\cdot|X^{(s)}, Y^{(s)})$

(b) compute $\gamma_X(X^*|X^{(s)}, Y^{(s)}) = \frac{f(X^*, Y^{(s)})}{f(X^{(s)}, Y^{(s)})} \cdot \frac{q_X(X^{(s)}|X^*, Y^{(s)})}{q_X(X^*|X^{(s)}, Y^{(s)})}$

(c) set $X^{(s+1)}$ to X^* with probability $\min(1, \gamma_X)$; set $X^{(s+1)}$ to $X^{(s)}$ otherwise.

2. Update Y

(a) sample $Y^* \sim q_Y(\cdot|X^{(s+1)}, Y^{(s)})$

(b) compute $\gamma_Y(Y^*|X^{(s+1)}, Y^{(s)}) = \frac{f(X^{(s+1)}, Y^*)}{f(X^{(s+1)}, Y^{(s)})} \cdot \frac{q_Y(Y^{(s)}|X^{(s+1)}, Y^*)}{q_Y(Y^*|X^{(s+1)}, Y^{(s)})}$

(c) set $Y^{(s+1)}$ to Y^* with probability $\min(1, \gamma_Y)$; set $Y^{(s+1)}$ to $Y^{(s)}$ otherwise.

Remark: $p((x, y) \rightarrow (x^*, y^*))f(x, y) = p((x^*, y^*) \rightarrow (x, y))f(x^*, y^*)$ does not hold for this algorithm in general.

Proof:

Extra: Multivariate Normal (for R code example in class)

The multivariate normal distribution of a d -dimensional random vector $\mathbf{X} = (X_1, \dots, X_d)^\top \in \mathbb{R}^d$ can be written in the notation $\mathbf{X} \sim \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$.

When the symmetric covariance matrix $\boldsymbol{\Sigma}$ is positive definite, then the multivariate normal distribution is non-degenerate, and the distribution has density function

$$f_{\mathbf{X}}(\mathbf{x}|\boldsymbol{\mu}, \boldsymbol{\Sigma}) = \frac{1}{(2\pi)^{\frac{d}{2}}} (\det(\boldsymbol{\Sigma}))^{-\frac{1}{2}} \exp\left(-\frac{1}{2}(\mathbf{x} - \boldsymbol{\mu})^\top \boldsymbol{\Sigma}^{-1}(\mathbf{x} - \boldsymbol{\mu})\right) \\ \propto \exp\left(-\frac{1}{2}\mathbf{x}^\top \boldsymbol{\Sigma}^{-1}\mathbf{x} + \mathbf{x}^\top \boldsymbol{\Sigma}^{-1}\boldsymbol{\mu}\right)$$

where $\mathbf{x} = (x_1, \dots, x_d)^\top$.

For comparison, the density function of (univariate) normal distribution $X \sim N(\mu, \sigma^2)$

$$f_X(x|\mu, \sigma^2) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right) \propto \exp\left(-\frac{1}{2\sigma^2}(x^2 - 2x\mu + \mu^2)\right) \propto \exp\left(-\frac{1}{2}x^2(\sigma^2)^{-1} + x(\sigma^2)^{-1}\mu\right)$$