

An Estimate for the Variance of MCMC Estimates ($\hat{\mu}_n$).

Let $\underline{X}_1, \underline{X}_2, \dots, \underline{X}_n, \dots$ be a Markov chain with invariant distribution π .

Suppose we wish to estimate

$$\mu = \int_S g(\underline{x}) d\pi(\underline{x})$$

Under regularity assumptions [e.g. the Mchain is irreducible and $E_{\pi} g^2(\underline{x}) < \infty$]

we have that

$$\hat{\mu}_n = \frac{1}{n} \sum_{i=1}^n g(\underline{X}_i) \longrightarrow \mu \quad \text{a.s.} [\pi]$$

and

$\sqrt{n} (\hat{\mu}_n - \mu) \rightarrow_d N(0, \sigma^2)$ where

$$\sigma^2 = \text{Var}_{\pi}(g(\underline{x}_1)) + 2 \sum_{k=1}^{\infty} \text{Cov}_{\pi}(g(\underline{x}_1), g(\underline{x}_{1+k}))$$

So ,

$$\hat{\mu}_n \underset{\text{appr.}}{\sim} N\left(\mu, \frac{\sigma^2}{n}\right)$$

But the naive estimator :

$$\sqrt{\frac{1}{n(n-1)} \sum_{i=1}^n [g(\underline{x}_i) - \hat{\mu}_n]^2}$$

will not be a good estimate for the standard error of $\hat{\mu}_n$. Usually the variables $g(\underline{x}_1)$ and $g(\underline{x}_{1+k})$ are positively correlated.

The Batch Estimator for the $SE(\hat{\mu}_n)$

Divide the chain in batches

$$\underline{X}_1, \dots, \underline{X}_b \longrightarrow \tilde{\mu}_1(b)$$

$$\underline{X}_{b+1}, \dots, \underline{X}_{2b} \longrightarrow \tilde{\mu}_2(b)$$

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$$\underline{X}_{(k-1)b+1}, \dots, \underline{X}_{kb} \longrightarrow \tilde{\mu}_k(b)$$

Clearly,

$$\hat{\mu}_n = \frac{1}{k} \sum \tilde{\mu}_i(b)$$

Moreover, if  $b$  is large enough

$$\sqrt{b} (\tilde{\mu}_i(b) - \mu) \sim N(0, \sigma^2)$$

$$\Rightarrow \tilde{\mu}_i(b) \underset{\text{approx}}{\sim} N\left(\mu, \frac{\sigma^2}{b}\right)$$

Therefore

$$\frac{1}{(k-1)} \sum_{i=1}^k \left( \tilde{\mu}_i(b) - \hat{\mu}_n \right)^2 \text{ estimates } \frac{\sigma^2}{b}$$

and

$$\frac{1}{(k-1)k} \sum_{i=1}^k \left( \tilde{\mu}_i(b) - \hat{\mu}_n \right)^2 \text{ estimates } \frac{\sigma^2}{bk} = \boxed{\frac{\sigma^2}{n}}$$

Note: we are assuming that the batch means,  $\tilde{\mu}_i(b)$  are approximately uncorrelated.

This will be the case for "b" large enough.

Hence, the Standard Error estimate for  $\hat{\mu}_n$  is

$$SD(\hat{\mu}_n) = \sqrt{\frac{1}{k(k-1)} \sum_{i=1}^k \left[ \tilde{\mu}_i(b) - \hat{\mu}_n \right]^2}$$

Note: the autocorrelation of the  $\tilde{\mu}_i(b)$  are computed for increasing values of b until it drops below some given  $\epsilon \approx 0.05$  or  $0.01$ .