

Gibbs Sampling

A Markov Chain algorithm to sample from the joint distribution $f_{\mathbf{X}}$ of a p -dimensional vector of random variables $\mathbf{X} = (X_1, \dots, X_p)$.

1. Specify an initial value $\mathbf{X}^{(0)} = (X_1^{(0)}, \dots, X_p^{(0)})$
2. Repeat for $j = 1, 2, \dots$,
 - (a) Generate $X_1^{(j+1)}$ from $f(X_1 | X_2^{(j)}, X_3^{(j)}, \dots, X_p^{(j)})$
 - (b) Generate $X_2^{(j+1)}$ from $f(X_2 | X_1^{(j+1)}, X_3^{(j)}, \dots, X_p^{(j)})$
 - (c) $\vdots$
 - (d) Generate $X_p^{(j+1)}$ from $f(X_p | X_1^{(j+1)}, \dots, X_{p-1}^{(j+1)})$.
3. Return the values $\mathbf{X}^{(1)}, \mathbf{X}^{(2)}, \mathbf{X}^{(3)}, \dots$.

This algorithm is a markov chain. The values that the algorithm generates are realisations of the random vector $\mathbf{X}$. The state space of the MC is therefore the set of values that $\mathbf{X}$ can take.

After running the MC for a while, it will converge and the values of $\mathbf{X}$ generated will be a random sample from the limiting distribution $f_{\mathbf{X}}$.