

A supplementary material for Lecture 8

Tomonari SEI

June 8, 2017

In this note, we prove the validity of the Gibbs sampler (Example 6.14.6 of PRP).

1 Description of the Gibbs sampler and its validity

Let $\Theta = S^V$, where S is a finite (or countable) set called the ‘local state space’ and V is a finite set called the ‘index set’. Suppose that we want to generate random samples from

$$\pi_i = \frac{\pi_i^*}{Z}, \quad i = (i_v)_{v \in V} \in \Theta,$$

where π_i^* is a given positive function^{*1} and $Z = \sum_{i \in \Theta} \pi_i^* < \infty$.

1.1 The Gibbs sampler for a fixed index

Fix an index $v \in V$. For each $i \in \Theta$, let

$$\Theta_{i,v} = \{j \in \Theta : j_w = i_w \text{ for } w \neq v\}.$$

Define

$$h_{ij} = \frac{\pi_j}{\sum_{k \in \Theta_{i,v}} \pi_k} = \frac{\pi_j^*}{\sum_{k \in \Theta_{i,v}} \pi_k^*}, \quad j \in \Theta_{i,v}.$$

The *Gibbs sampler* for the fixed index v is a Markov chain $\{X_n\}_{n \geq 0}$ described as follows.

- If the current state is $X_n = i$, then $X_{n+1} = j$ is generated according to the probability mass function $(h_{ij})_{j \in \Theta_{i,v}}$.

Denote the transition matrix of the Markov chain by $\mathbf{P}^v = (p_{ij}^v)$. Note that \mathbf{P}^v is *not* irreducible (except for some trivial cases) since v is fixed.

Lemma 1. The transition matrix \mathbf{P}^v satisfies the detailed balance equation with respect to $\boldsymbol{\pi} = (\pi_i)_{i \in \Theta}$.

^{*1} If the function π_i^* is allowed to take the zero value, it will need additional care about irreducibility (see Lemma 2 and Lemma 3).

Proof. The transition matrix is given by $p_{ij}^v = h_{ij}$ if $j \in \Theta_{i,v}$, and $p_{ij} = 0$ otherwise. If $j \in \Theta_{i,v}$, then $\Theta_{j,v} = \Theta_{i,v}$ and therefore

$$\pi_i p_{ij}^v = \frac{\pi_i \pi_j}{\sum_{k \in \Theta_{i,v}} \pi_k} = \frac{\pi_i \pi_j}{\sum_{k \in \Theta_{j,v}} \pi_k} = \pi_j p_{ji}^v.$$

If $j \notin \Theta_{i,v}$, then $i \notin \Theta_{j,v}$ and therefore

$$\pi_i p_{ij}^v = 0 = \pi_j p_{ji}^v.$$

Hence the detailed balance equation holds. \square

It is necessary to modify the Markov chain in order to make it irreducible. As described in Example 6.14.6 of PRP, we may choose the index v at random or in pre-determined manner. We discuss the two methods in the following subsections.

1.2 The Gibbs sampler with the randomized index

If the index v is randomly chosen, the resultant chain has the transition matrix

$$\mathbf{P} = \frac{1}{|V|} \sum_{v \in V} \mathbf{P}^v.$$

Lemma 2. The matrix \mathbf{P} is irreducible and satisfies the detailed balance equation.

Proof. The detailed balance equation of \mathbf{P} follows from that of \mathbf{P}^v . To check the irreducibility, let i and j be any states in Θ . Assign numbers of the elements of V as $V = \{1, \dots, n\}$ without loss of generality. Define $i^{(0)}, i^{(1)}, \dots, i^{(n)}$ by

$$i_v^{(k)} = \begin{cases} j_v & \text{if } v \leq k, \\ i_v & \text{if } v > k. \end{cases}$$

for $0 \leq k \leq n$. Then we have $i^{(0)} = i$, $i^{(n)} = j$, and

$$p_{i^{(k-1)} i^{(k)}} \geq \frac{1}{|V|} p_{i^{(k-1)} i^{(k)}}^k > 0$$

for $1 \leq k \leq n$. Hence \mathbf{P} is irreducible. \square

1.3 The Gibbs sampler with the pre-determined index

Let $V = \{1, \dots, n\}$ without loss of generality. If we apply the Markov chain \mathbf{P}^v for $v = 1, \dots, n$ in order, then the resultant Markov chain is given by

$$\mathbf{P} = \mathbf{P}^1 \dots \mathbf{P}^n.$$

In general, \mathbf{P} does not satisfy the detailed balance equation. But the following lemma holds.

Lemma 3. The matrix \mathbf{P} is irreducible and has the stationary distribution $\boldsymbol{\pi}$.

Proof. Irreducibility is proved in the same manner as Lemma 2. For stationarity, we have

$$\begin{aligned}\boldsymbol{\pi}\mathbf{P} &= \boldsymbol{\pi}\mathbf{P}^1 \dots \mathbf{P}^n \\ &= \boldsymbol{\pi}\end{aligned}$$

since $\boldsymbol{\pi}$ is a stationary distribution of each \mathbf{P}^v . □