Markov Chains and Algorithmic Applications - IC - EPFL

## Solutions 5

1. "Only if" part: Let us fix n and the sequence of  $j_1, j_2, ..., j_n$ . By using the detailed balance equation, we have

$$\pi_{j_1} p_{j_1 j_2} p_{j_2 j_3} \cdots p_{j_{n-1} j_n} p_{j_n j_1} = p_{j_2 j_1} \pi_{j_2} p_{j_2 j_3} \cdots p_{j_{n-1} j_n} p_{j_n j_1}$$

$$= p_{j_2 j_1} p_{j_3 j_2} \pi_{j_3} \cdots p_{j_{n-1} j_n} p_{j_n j_1}$$

$$= \dots$$

$$= p_{j_2 j_1} p_{j_3 j_2} \cdots p_{j_n j_{n-1}} \pi_{j_n} p_{j_n j_1}$$

$$= p_{j_2 j_1} p_{j_3 j_2} \cdots p_{j_n j_{n-1}} p_{j_1 j_n} \pi_{j_1}$$

As we know the chain is ergodic,  $\pi_{j_1} \neq 0$ , so we have

$$p_{j_1j_2} p_{j_2j_3} \cdots p_{j_{n-1}j_n} p_{j_nj_1} = p_{j_2j_1} p_{j_3j_2} \cdots p_{j_nj_{n-1}} p_{j_1j_n}$$

"If" part: Let us fix n and marginalize the expression  $p_{j_1j_2} p_{j_2j_3} \cdots p_{j_{n-1}j_n} p_{j_nj_1}$  over the sequences of  $j_2, \ldots, j_{n-1}$  as

$$\sum_{j_2,\dots,j_{n-1}} p_{j_1j_2} p_{j_2j_3} \cdots p_{j_{n-1}j_n} p_{j_nj_1} = p_{j_nj_1} \sum_{j_2,\dots,j_{n-1}} p_{j_1j_2} p_{j_2j_3} \cdots p_{j_{n-1}j_n}$$
$$= p_{j_nj_1} \sum_{j_2,\dots,j_{n-1}} \mathbb{P}(X_n = j_n, X_{n-1} = j_{n-1},\dots, X_2 = j_2 | X_1 = j_1)$$
$$= p_{j_nj_1} \mathbb{P}(X_n = j_n | X_1 = j_1) = p_{j_nj_1} p_{j_1j_n}^{(n-1)}$$

By repeating the same set of calculations for the expression  $p_{j_1j_n} p_{j_nj_{n-1}} \cdots p_{j_3j_2} p_{j_2j_1}$ , using the assumed equality, and considering the case for which  $j_1 = i$  and  $j_n = k$  (i.e., fixing the first and the last state in the sequence), we have

$$p_{ki} \, p_{ik}^{(n-1)} = p_{ik} \, p_{ki}^{(n-1)}$$

Since the chain is ergodic, as n goes to  $+\infty$ ,  $p_{ik}^{(n-1)}$  and  $p_{ki}^{(n-1)}$  go to  $\pi_k$  and  $\pi_i$  respectively. Therefore, we have

$$p_{ki}\pi_k = p_{ik}\pi_i$$

which is the detailed balance equation.

2. a) This chain is clearly ergodic. The transition matrix is

$$\begin{pmatrix} 1-p & p & 0 \\ 1/2 & 0 & 1/2 \\ 0 & p & 1-p \end{pmatrix}$$

Assume that the detailed balance equation is satisfied. Then

$$\pi_1^*/2 = \pi_2^* p = \pi_0^* p$$

We conclude that

$$\pi_0^* = \pi_2^* = \frac{1}{2(1+p)}$$
  $\pi_1^* = \frac{p}{1+p}$ 

It is then easy to verify that  $\pi^* = \pi^* P$ , and so this is indeed a stationary distribution, which obviously satisfies the detailed balance equation.

**b**) We know that  $\lambda_0 = 1$ , and so, to compute the eigenvalues, we must solve the equations

$$2 - 2p = 1 + \lambda_1 + \lambda_2$$
$$-p(1 - p) = \lambda_1 \lambda_2$$

Solving this, we obtain that  $\lambda_1 = 1 - p$  and  $\lambda_2 = -p$ . So  $\lambda_* = \max(p, 1 - p)$  and the spectral gap is given by  $\gamma = 1 - \lambda_* = \min(p, 1 - p)$ .

c) For  $p = \frac{1}{N}$ , the spectral gap is  $\gamma = \frac{1}{N}$ . From the theorem seen in class, we know that  $||P_i^n - \pi||_{\text{TV}} \leq \frac{\exp(-\gamma n)}{2\sqrt{\pi_i}}$ , so here,

$$\max_{i \in S} \|P_i^n - \pi\|_{\mathrm{TV}} \le \frac{1}{2} \sqrt{\frac{1 + 1/N}{1/N}} \exp(-n/N) \le \sqrt{N} \exp(-n/N) = \exp\left(\frac{\log N}{2} - \frac{n}{N}\right)$$

Taking therefore  $n \ge N\left(\frac{\log N}{2} + c\right)$  with c > 0 sufficiently large (more precisely,  $c = \log(1/\varepsilon)$ ) ensures that the maximum total variation norm is below  $\varepsilon$ .

**d**) For  $p = 1 - \frac{1}{N}$ , the spectral gap is again  $\gamma = \frac{1}{N}$ . So

$$\max_{i \in S} \|P_i^n - \pi\|_{\mathrm{TV}} \le \frac{1}{2} \sqrt{2(2 - 1/N)} \exp(-n/N) \le \exp(-n/N)$$

Taking therefore  $n \ge cN$  with  $c = \log(1/\varepsilon)$  ensures that the maximum total variation norm is below  $\varepsilon$ , so

$$T_{\varepsilon} \leq N \log(1/\varepsilon)$$

**3.** a) The transition matrix being doubly stochastic, the stationary distribution is uniform (i.e.  $\pi_i = \frac{1}{2N}$  for every  $i \in S$ ) and satisfies the detailed balance equation.

**b)** Solving the equation  $P\phi^{(1)} = \lambda \phi^{(1)}$ , we obtain

$$\frac{N-1}{N}a + \frac{1}{N}b = \lambda a$$
$$\frac{N-1}{N}a - \frac{1}{N}b = \lambda b$$

which is saying that  $\lambda$  is an eigenvalue of the 2  $\times$  2 matrix

$$\begin{pmatrix} \frac{N-1}{N} & \frac{1}{N} \\ \frac{N-1}{N} & -\frac{1}{N} \end{pmatrix} = \begin{pmatrix} 1-\delta & \delta \\ 1-\delta & -\delta \end{pmatrix}$$

where we have set  $\delta = \frac{1}{N}$ . These eigenvalues are given by

$$\lambda_{\pm} = \frac{1 - 2\delta \pm \sqrt{(1 - 2\delta)^2 + 8\delta(1 - \delta)}}{2} = \frac{1 - 2\delta \pm \sqrt{1 + 4\delta - 4\delta^2}}{2}$$

For  $\delta$  small (i.e. N large), the largest of these 2 eigenvalues is  $\lambda_+$ , which is approximately given by

$$\lambda_{+} \simeq \frac{1 - 2\delta + (1 + 2\delta - 4\delta^{2})}{2} = 1 - 2\delta^{2} = 1 - \frac{2}{N^{2}}$$

so the spectral gap  $\gamma \simeq \frac{2}{N^2}$ .

c) By the theorem seen in class,

$$\max_{i \in S} \|P_i^n - \pi\|_{\text{TV}} \le \frac{\sqrt{2N}}{2} \exp(-\gamma n) \le \sqrt{2} \exp\left(\frac{\log N}{2} - \frac{2n}{N^2}\right)$$

is below  $\varepsilon$  for  $n \ge \frac{N^2}{2} \left( \frac{\log N}{2} + \log(\sqrt{2}/\varepsilon) \right)$ , so

$$T_{\varepsilon} \leq \frac{N^2}{2} \left( \frac{\log N}{2} + \log(\sqrt{2}/\varepsilon) \right)$$