

### Exam of 12.10.2007 — Solutions

**1** \_\_\_\_\_

**1a** .....

The absorbing states 0 and  $N/8$  are reachable from every other state, therefore all other states are transient. A transient state is visited only a finite number of times (almost surely); the same holds for the finite set of transient states.

**1b** .....

$p(x, x - 1) = x/N$ , thus

$$\mathbb{P}(B) = p(i, i - 1)p(i - 1, i - 2) \dots p(1, 0) = \frac{i}{N} \frac{i - 1}{N} \dots \frac{1}{N} = \frac{i!}{N^i}.$$

**1c** .....

We have to check that  $h(x) \geq p(x, x - 1)h(x - 1) + p(x, x + 1)h(x + 1)$  for  $0 < x < N/8$ ; that is,

$$x!(N - x)! \left(1 + \frac{4x}{N}\right) \geq \frac{x}{N}(x - 1)!(N - x + 1)! \left(1 + \frac{4(x - 1)}{N}\right) + \frac{N - x}{N}(x + 1)!(N - x - 1)! \left(1 + \frac{4(x + 1)}{N}\right).$$

We cancel  $x!(N - x)!$  and simplify:

$$\begin{aligned} 1 + \frac{4x}{N} &\geq \frac{N - x + 1}{N} \left(1 + \frac{4x - 4}{N}\right) + \frac{x + 1}{N} \left(1 + \frac{4x + 4}{N}\right); \\ 1 + \frac{4x}{N} &\geq \frac{N + 2}{N} \left(1 + \frac{4x}{N}\right) - \frac{N - 2x}{N} \cdot \frac{4}{N}; \\ \frac{N - 2x}{N} \cdot \frac{4}{N} &\geq \frac{2}{N} \left(1 + \frac{4x}{N}\right); \\ \frac{N - 2x}{N} \cdot 2 &\geq 1 + \frac{4x}{N}; \quad 2N - 4x \geq N + 4x; \quad N \geq 8x, \end{aligned}$$

which is true.

**1d** .....

By the stopping theorem for bounded supermartingales,

$$\mathbb{E} M_0 \geq \mathbb{E} M_T,$$

where  $T$  is the first time  $X_n$  visits 0 or  $N/8$ . We have

$$\begin{aligned} \mathbb{E} M_0 &= h(i), \\ \mathbb{E} M_T &= h(0)\mathbb{P}(A) + h(N/8)(1 - \mathbb{P}(A)) \geq h(0)\mathbb{P}(A). \end{aligned}$$

Therefore  $h(i) \geq h(0)\mathbb{P}(A)$ ;

$$\mathbb{P}(A) \leq \frac{h(i)}{h(0)} = \frac{i!(N-i)!(1 + \frac{4i}{N})}{N!} = \frac{1 + \frac{4i}{N}}{\binom{N}{i}}.$$

**1e** .....

Yes,  $\mathbb{P}(B|A)$  converges to 1 as  $N \rightarrow \infty$ , since

$$\mathbb{P}(B|A) = \frac{\mathbb{P}(B)}{\mathbb{P}(A)} \geq \frac{i!}{N^i} \cdot \frac{\binom{N}{i}}{1 + \frac{4i}{N}} = \frac{N(N-1)\dots(N-i+1)}{N^i} \cdot \frac{1}{1 + \frac{4i}{N}} \rightarrow 1.$$

**2** \_\_\_\_\_

**2a** .....

We have jump rates  $q(1, 2) = \lambda_1$ ,  $q(2, 1) = \lambda_2$ . The stationary distribution  $\pi(1), \pi(2)$  satisfies  $\pi_1 q(1, 2) = \pi_2 q(2, 1)$ , that is,  $\pi_1 \lambda_1 = \pi_2 \lambda_2$ . Thus,

$$\frac{\pi_1}{1/\lambda_1} = \frac{\pi_2}{1/\lambda_2} \quad \text{and} \quad \pi_1 + \pi_2 = 1;$$

$$\pi_1 = \frac{\frac{1}{\lambda_1}}{\frac{1}{\lambda_1} + \frac{1}{\lambda_2}} = \frac{\lambda_2}{\lambda_1 \lambda_2}.$$

**2b** .....

Similarly,

$$q(1, 2) = \lambda_1, \quad q(2, 3) = \lambda_2, \quad q(3, 1) = \lambda_3;$$

$$\pi_1 \lambda_1 = \pi_2 \lambda_2 = \pi_3 \lambda_3; \quad \pi_k = \frac{\frac{1}{\lambda_k}}{\frac{1}{\lambda_1} + \frac{1}{\lambda_2} + \frac{1}{\lambda_3}};$$

$$\pi_1 = \frac{\lambda_2 \lambda_3}{\lambda_2 \lambda_3 + \lambda_3 \lambda_1 + \lambda_1 \lambda_2}, \quad \pi_2 = \frac{\lambda_3 \lambda_1}{\lambda_2 \lambda_3 + \lambda_3 \lambda_1 + \lambda_1 \lambda_2}, \quad \pi_3 = \frac{\lambda_1 \lambda_2}{\lambda_2 \lambda_3 + \lambda_3 \lambda_1 + \lambda_1 \lambda_2}.$$

**2c** .....

Now  $q(1, 2) = q(1, 3) = \lambda_1/2$ ,  $q(2, 3) = q(2, 1) = \lambda_2/2$ ,  $q(3, 1) = q(3, 2) = \lambda_3/2$ . Stationarity:  $\pi_1 \lambda_1 = (\pi_2 \lambda_2 + \pi_3 \lambda_3)/2$ ,  $\pi_2 \lambda_2 = (\pi_3 \lambda_3 + \pi_1 \lambda_1)/2$ ,  $\pi_3 \lambda_3 = (\pi_1 \lambda_1 + \pi_2 \lambda_2)/2$ . It follows that  $\pi_1 \lambda_1 = \pi_2 \lambda_2 = \pi_3 \lambda_3$ , which can be shown algebraically or just by noting that the greatest among the three numbers  $\pi_k \lambda_k$  is strictly larger than the mean of the other two numbers, unless they all are equal. Thus,  $\pi_1, \pi_2, \pi_3$  are the same as in the previous case.

**2d** .....

This time  $q(1, 2) = \lambda_1$ ,  $q(2, 3) = \lambda_2$  and  $q(3, 1) = q(3, 2) = \lambda_3/2$ . Stationarity:  $\pi_1\lambda_1 = \pi_3\lambda_3/2$ ,  $\pi_2\lambda_2 = \pi_1\lambda_1 + 0.5\pi_3\lambda_3$ ,  $\pi_3\lambda_3 = \pi_2\lambda_2$ . Thus,  $0.5\pi_2\lambda_2 = \pi_1\lambda_1$ ;

$$\frac{\pi_1}{1/\lambda_1} = \frac{\pi_2}{2/\lambda_2} = \frac{\pi_3}{2/\lambda_3} \quad \text{and} \quad \pi_1 + \pi_2 + \pi_3 = 1;$$

$$\pi_1 = \frac{\frac{1}{\lambda_1}}{\frac{1}{\lambda_1} + \frac{2}{\lambda_2} + \frac{2}{\lambda_3}} = \frac{\lambda_2\lambda_3}{\lambda_2\lambda_3 + 2\lambda_3\lambda_1 + 2\lambda_1\lambda_2},$$

$$\pi_2 = \frac{\frac{2}{\lambda_2}}{\frac{1}{\lambda_1} + \frac{2}{\lambda_2} + \frac{2}{\lambda_3}} = \frac{2\lambda_1\lambda_3}{\lambda_2\lambda_3 + 2\lambda_3\lambda_1 + 2\lambda_1\lambda_2},$$

$$\pi_3 = \frac{\frac{2}{\lambda_3}}{\frac{1}{\lambda_1} + \frac{2}{\lambda_2} + \frac{2}{\lambda_3}} = \frac{2\lambda_1\lambda_2}{\lambda_2\lambda_3 + 2\lambda_3\lambda_1 + 2\lambda_1\lambda_2}.$$

**3** \_\_\_\_\_

**3a** .....

Yes, it can happen. For example, let  $X_n$  be a Markov chain with three values  $-1, 0, +1$  and transition probabilities  $p(-1, 0) = p(0, 1) = p(1, -1) = 1$ . Then the process  $Y_n = X_n^2$  is not Markov. Indeed,  $\mathbb{P}(Y_2 = 0 | Y_1 = 1, Y_0 = 1) = 1$  but  $\mathbb{P}(Y_2 = 0 | Y_1 = 1, Y_0 = 0) = 0$ .

**3b** .....

Yes, it can happen. For example, let  $X_n$  be a non-Markov process with two values  $-1, +1$  (such processes exist, see the previous item). Then the process  $Y_n = X_n^2$ , taking only one value  $+1$ , is Markov.

**3c** .....

No, it cannot happen. Markovianity of the process  $X_n$  is equivalent to Markovianity of the process  $Y_n = X_n^3$ , since the function  $y = x^3$  is one-to-one.