## Probability III - 2007/08

## **Solutions to Exercise Sheet 4**

1. Let the state space for the chain be  $\{1, 2, 3, 4, 5, 6\}$ . From the transition matrix it can be seen that the equivalence classes are:

$$\{1\},\{2,5\},\{3,4\},\{6\}.$$

If you're having trouble seeing this then try drawing the tranistion graph.

States 1 and 6 are absorbing so are certainly recurrent. States 2 and 5 are also recurrent (the probability of not returning to state 2 by *n* steps is  $\frac{1^n}{2} \rightarrow 0$ ). Finally, states 3 and 4 are transient (the probability that we leave  $\{3,4\}$  eventually is 1 and having left we can never return).

2. The probability that of the first *n* steps, *k* of them were to the right is  $2^{-n} \binom{n}{k}$ . If we return to 0 after *n* steps then we must have made exactly twice as many to the left as to the right. Thus

$$p_{00}^{(n)} = \begin{cases} 2^{-n} {n \choose \frac{n}{3}} & \text{if } n \text{ is a multiple of 3} \\ 0 & \text{otherwise} \end{cases}$$

Now,

$$\sum_{k\geq 0} p_{00}^{(k)} = \sum_{k\geq 0} 2^{-3k} \binom{3k}{k}$$
$$= \sum_{k\geq 0} 2^{-3k} \frac{(3k)!}{k!(2k)!}$$
$$\leq \sum_{k\geq 0} 2^{-3k} C \frac{(3k)^{(3k)} \sqrt{3k} e^k e^{2k}}{e^{3k} k^k \sqrt{k}(2k)^{2k} \sqrt{2k}} \quad \text{(for some constant } C\text{)}$$
$$= \sum_{k\geq 0} C \frac{3^{3k} \sqrt{3}}{2^{3k} 2^{2k} \sqrt{k}}$$
$$\leq \sqrt{3} \sum_{k\geq 0} \frac{27^k}{32^k}$$
$$= \frac{5\sqrt{3}}{32} \quad \text{(sum of a geometric progression)}$$

We deduce that since  $\sum_{k\geq 0} p_{00}^{(k)}$  is finite, state 0 is transient. Since the chain is irreducible this means that every state is transient.

3. Using the notation of the lectures we have that

$$p_k = (1-p)p^k,$$

and hence,

$$g(z) = \sum_{k \ge 0} p_k z^k$$
$$= \frac{1-p}{1-pz}.$$

You might have noticed that the  $p_k$  have geometric distribution, so g(z) is the generating function for a geometric random variable.

We know that *u*, the probability of eventual extinction, is 1 if  $g'(1) \le 1$ . Differentiating *g* gives,

$$g'(z) = \frac{p(1-p)}{(1-pz)^2}$$

Hence,  $g'(1) = \frac{p}{1-p}$ . This is  $\leq 1$  if  $p \leq \frac{1}{2}$ .

If  $p < \frac{1}{2}$  then g'(1) > 1 and so we know from the general theory that g(z) = z has two roots, 1 and a smaller root in the interval (0, 1). We also know that *u* is equal to the smaller root. So to find *u* we need to solve

$$u = \frac{1-p}{1-pu}.$$

That is,

$$pu^{2} - u + 1 - p = 0$$
$$(pu - (1 - p))(u - 1) = 0$$

Hence  $u = \frac{1-p}{p}$ . So if  $p < \frac{1}{2}$  then the probability of eventual extinction is  $\frac{1-p}{p}$ . By the general theory developed in lectures we know that  $G_3(z)$ , the generating function for  $X_3$ , satisfies

$$G_3(z) = g(g(g(z)))$$

When  $p = \frac{1}{3}$ , we have that  $g(z) = \frac{2}{3-z}$ . Thus,

$$G_3(z) = g\left(g\left(\frac{2}{3-z}\right)\right)$$
$$= g\left(\frac{2}{3-\frac{2}{3-z}}\right)$$
$$= g\left(\frac{6-2z}{7-3z}\right)$$
$$= \frac{2}{3-\frac{6-2z}{7-3z}}$$
$$= \frac{14-6z}{15-7z}.$$

4. When reading these solutions, make sure that you can see which properties of the Poisson process we are using to justify each step below.

a)

$$\mathbb{P}(X(4) = 2) = \mathbb{P}(X(4) - X(0) = 2)$$
$$= e^{-4 \times 4} \frac{(4 \times 4)^2}{2!}$$
$$= 128e^{-16}.$$

b)

$$\mathbb{P}(X(4) = 2, X(1) = 1) = \mathbb{P}(X(1) - X(0) = 1)\mathbb{P}(X(4) - X(1) = 1)$$
$$= e^{-4}4e^{-12}12$$
$$= 48e^{-16}.$$

c)

$$\mathbb{P}(X(4) = 2|X(1) = 1) = \mathbb{P}(X(4) - X(1) = 1)$$
  
=  $12e^{-12}$ .

d)

$$\mathbb{P}(X(1) = 1 | X(4) = 2) = \frac{\mathbb{P}(X(1) = 1, X(4) = 2)}{\mathbb{P}(X(4) = 2)}$$
$$= \frac{48e^{-16}}{128e^{-16}}$$
$$= \frac{3}{8}.$$

e)

$$\mathbb{P}(X(4) = 1 | X(1) = 2) = 0.$$

5. Let X(t) be the number of cracks in the first *t* miles and Y(t) be the number of punctures in the first *t* miles. We must check that Z(t) = X(t) + Y(t) is a Poisson process.

Firstly, it is clear that Z(0) = 0 (since the fact that *X* and *Y* are Poisson processes means that X(0) = Y(0) = 0). Secondly, we have that for any  $t_1 < t_2 < \cdots < t_n$ ,  $X(t_{i+1}) - X(t_i)$ ,  $Y(t_{i+1}) - Y(t_i)$  are independent random variables. Hence,  $Z(t_{i+1}) - Z(t_i)$  are independent random variables. Finally, we must show that Z(t+s) - Z(s) has a Po $(t(\lambda + \mu))$  distribution. Now,

$$Z(t+s) - Z(s) = X(t+s) - X(s) + Y(t+s) - Y(s).$$

We know that X(t+s) - X(s) is distributed  $Po(t\lambda)$  and Y(t+s) - Y(s) is distributed  $Po(t\mu)$ . Moreover, these two random variables are independent. It follows (by standard properties of the Poisson distribution – see for example exercise sheet 1 question 1 (b)) that their sum is distributed  $Po(t(\lambda + \mu))$ , as required.

Let F(t) be the number of un-repairable faults ni the first *t* miles. Again, it is immediate that F(0) = 0, and that for any  $t_1 < t_2 < \cdots < t_n$ ,  $F(t_{i+1}) - F(t_i)$  are independent random variables. By standard properties of the Poisson distribution (see for example exercise sheet 1 question 3) we have that F(t+s) - F(s) is distributed  $Po(p\lambda)$  as required.

## Please let me know if you have any comments or corrections