

**Practica 5. Cadenas de Markov**  
Procesos Estocásticos 2015.

1. Pruebe que si  $\mu$  es reversible para  $Q$ , entonces  $\mu$  es invariante para  $Q$ .
2. Pruebe que si  $\mu$  es invariante para  $Q$ , entonces la matriz  $Q^*$  definida por

$$Q^*(x, y) = \frac{\mu(y)}{\mu(x)} Q(y, x). \tag{1}$$

es una matriz de transición (filas suman 1)

3. Sabemos que la probabilidad que un ciclo sea recorrido en el sentido directo por el proceso  $X_n$  es igual a la probabilidad que sea recorrido en el sentido reverso por  $X_n^*$ . Use ese resultado para demostrar que  $T^{a,a}$  y  $(T^*)^{a,a}$  tienen la misma distribución.
4. Calcule las medidas invariantes para las cadenas de Markov de los ejemplos de las teóricas.
5. Pruebe que la medida  $\mu$  definida al discutir el paseo asimétrico en el círculo es invariante para  $Q$  y calcule la matriz reversa  $Q^*$  en relación a  $\mu$ . Este es un ejemplo de una cadena de Markov que tiene una medida invariante que no es reversible.
6. Sea  $Q$  una matriz irreducible en un espacio finito  $G$ . Sea

$$\mu(y) := \frac{1}{E(T^{x \rightarrow x})} \sum_{n \geq 0} P(X_n^x = y, T^{x \rightarrow x} > n) \tag{2}$$

O sea,  $\mu(y)$  es proporcional al número esperado de visitas al estado  $y$  antes que la cadena vuelva a  $x$ . Muestre que  $\mu$  es invariante para  $Q$ . Esta es una prueba alternativa para el Teorema de Perron Frobenius.

7. Birth and death process The birth and death process is a Markov chain on  $\mathbb{N}$  with transition matrix:

$$Q(x, x + 1) = q(x) = 1 - Q(x, x - 1), \text{ if } x \geq 1, \\ Q(0, 1) = q(0) = 1 - Q(0, 0),$$

where  $(q(x))_{x \in \mathbb{N}}$  is a sequence of real numbers contained in  $(0, 1)$ . Under which condition on  $(q(x))_{x \in \mathbb{N}}$  the process accepts an invariant measure? Specialize to the case  $q(x) = p$  for all  $x \in \mathbb{N}$ . Pista: plantee las ecuaciones de balance y busque soluciones normalizables.

8. Método de Monte Carlo One of the most popular Monte Carlo methods to obtain samples of a given probability measure consists in simulate a Markov chain having the target measure as invariant measure. To obtain a sample of the target measure from the trajectory of the Markov chain, one needs to let the process evolve until it “attains equilibrium”. We propose a chain for the simulation.

i) Let  $N$  be a positive integer and  $\mu$  a probability measure on  $\{1, \dots, N\}$ . Consider the following transition probabilities on  $\{1, \dots, N\}$ : for  $y \neq x$ ,

$$Q_1(x, y) = \frac{1}{N - 1} \frac{\mu(y)}{\mu(x) + \mu(y)} \tag{3}$$

$$Q_1(x, x) = 1 - \sum_{y \neq x} Q_1(x, y) \tag{4}$$

(Choose uniformly a state  $y$  different of  $x$  and with probability  $\mu(y)/(\mu(x) + \mu(y))$  jump to  $y$ ; with the complementary probability stay in  $x$ .)

$$Q_2(x, y) = \frac{1}{N-1} \left[ \mathbf{1}\{\mu(x) \leq \mu(y)\} + \frac{\mu(y)}{\mu(x)} \mathbf{1}\{\mu(x) > \mu(y)\} \right] \quad (5)$$

$$Q_2(x, x) = 1 - \sum_{y \neq x} Q_2(x, y). \quad (6)$$

(Choose uniformly a state  $y$  different of  $x$  and if  $\mu(x) < \mu(y)$ , then jump to  $y$ ; if  $\mu(x) \geq \mu(y)$ , then jump to  $y$  with probability  $\mu(y)/\mu(x)$ ; with the complementary probability stay in  $x$ .)

Check if  $\mu$  is reversible with respect to  $Q_1$  and/or  $Q_2$ .

ii) Let  $G = \{0, 1\}^N$  and  $\mu$  be the uniform distribution on  $G$  (that is,  $\mu(\zeta) = 2^{-N}$ , for all  $\zeta \in G$ ). Define a transition matrix  $Q$  on  $G$  satisfying the following conditions:

a)  $Q(\xi, \zeta) = 0$ , if  $d(\xi, \zeta) \geq 2$ , where  $d$  is the Hamming's distance, introduced in the previous chapter;

b)  $\mu$  is reversible with respect to  $Q$ .

iii) Simulate the Markov chain corresponding to the chosen matrix  $Q$ . How would you determine empirically the moment when the process attains equilibrium? Hint: plot the relative frequency of visit to each site against time and wait this to stabilize. Give an empiric estimate of the density of ones. Compare with the true value given by  $\mu(1)$ , where  $\mu$  is the invariant measure for the chain.

iv) Use the Ehrenfest model to simulate the Binomial distribution with parameters  $\frac{1}{2}$  and  $N$ .

9. (i) Prove that if  $Q$  is the matrix

$$Q = \begin{pmatrix} p & 1-p \\ 1-q & q \end{pmatrix} \quad (7)$$

then  $Q^n$  converges to the matrix

$$\begin{pmatrix} \frac{1-q}{(1-p) + (1-q)} & \frac{1-p}{(1-p) + (1-q)} \\ \frac{1-q}{(1-p) + (1-q)} & \frac{1-p}{(1-p) + (1-q)} \end{pmatrix} \quad (8)$$

ii) For the same chain compute  $E(T^{1 \rightarrow 1})$ , where

$$T^{1 \rightarrow 1} = \inf\{n \geq 1 : X_n^1 = 1\}.$$

iii) Establish a relationship between items (i) and (ii).

10. Let  $G = \mathbb{N}$  and  $Q$  be a transition matrix defined as follows. For all  $x \in \mathbb{N}$

$$Q(0, x) = p(x) \text{ and}$$

$$Q(x, x-1) = 1 \text{ if } x \geq 1,$$

where  $p$  is a probability measure on  $\mathbb{N}$ . Let  $(X_n^0)_{n \in \mathbb{N}}$  be a Markov chain with transition matrix  $Q$  and initial state 0.

i) Give sufficient conditions on  $p$  to guarantee that  $Q$  has at least one invariant measure.

ii) Compute  $E(T^{1 \rightarrow 1})$ .

iii) Establish a relationship between items (i) and (ii).

11. Stirling process The stirring process is a Markov chain on the hypercube  $G = \{0, 1\}^N$  defined by the following algorithm. Let  $\mathcal{P}_N$  be the set of all possible permutations of the sequence  $(1, 2, \dots, N)$ , that is, the set of bijections from  $\{1, 2, \dots, N\}$  into itself. Let  $\pi$  be an element of  $\mathcal{P}_N$ . Let  $F_\pi: G \rightarrow G$  be the function defined as follows. For all  $\xi \in G$

$$F_\pi(\xi)(i) = \xi(\pi(i)).$$

In other words,  $F_\pi$  permutes the values of each configuration  $\xi$  assigning to the coordinate  $i$  the former value of the coordinate  $\pi(i)$ .

Let  $(\Pi_1, \Pi_2, \dots)$  be a sequence of iid random variables on  $\mathcal{P}_N$ . The stirring process  $(\eta_n^\zeta)_{n \in \mathbb{N}}$  with initial state  $\zeta$  is defined as follows:

$$\eta_n^\zeta = \begin{cases} \zeta, & \text{if } n = 0; \\ F_{\Pi_n}(\eta_{n-1}^\zeta), & \text{if } n \geq 1 \end{cases} \quad (9)$$

- i) Show that the stirring process is not irreducible (it is *reducible!*).
- ii) Assume that the random variables  $\Pi_n$  have uniform distribution on  $\mathcal{P}_N$ . Which are all the invariant measures for the stirring process in this case?
- iii) Let  $\mathcal{V}_N$  be the set of permutations that only change the respective positions of two neighboring points of  $(1, 2, \dots, N)$ . A typical element of  $\mathcal{V}_N$  is the permutation  $\pi^k$ , for  $k \in \{1, 2, \dots, N\}$ , defined by:

$$\pi^k(i) = \begin{cases} i, & \text{if } i \neq k, i \neq k + 1 \\ k + 1, & \text{if } i = k, \\ k, & \text{if } i = k + 1. \end{cases} \quad (10)$$

In the above representation, the sum is done “module  $N$ ”, that is,  $N + 1 = 1$ . Assume that the random variables  $\Pi_n$  are uniformly distributed in the set  $\mathcal{V}_N$ . Compute the invariant measures of the stirring process in this case.

- iv) Compare the results of items (ii) and (iii).

12. Pruebe que si  $\mu$  es invariante para  $Q$ , entonces

$$\mu(b) = \sum_{x \in G} \mu(x) Q^k(x, b) \quad (11)$$

para todo  $k \geq 1$ .

13. Pruebe que los acoplamiento libre e independiente coinciden después del instante de encuentro:

$$n \geq \tau^{a,b} \quad \text{implica} \quad X_n^a = X_n^b \quad (12)$$

14. *Independent coalescing coupling.* Let  $Q$  be a transition matrix on the finite or countable set  $G$ . Define the matrix  $\bar{Q}$  on  $G \times G$  as follows

$$\bar{Q}((a, b), (x, y)) = \begin{cases} Q(a, x)Q(b, y), & \text{if } a \neq b; \\ Q(a, x), & \text{if } a = b \text{ and } x = y; \\ 0, & \text{if } a = b \text{ and } x \neq y. \end{cases} \quad (13)$$

Verify that  $\bar{Q}$  is a transition matrix. In other words, verify that for all  $(a, b) \in G \times G$ ,

$$\sum_{(x,y) \in G \times G} \bar{Q}((a, b), (x, y)) = 1.$$

Observe that the chain corresponding to  $\bar{Q}$  describes two Markov chains of transition matrix  $Q$  which evolve independently up to the first time both visit the same state. From this moment on, the two chains continue together for ever.

15. *Independent coalescing coupling.* Show that the process defined in Example ?? has transition matrix  $Q$  defined by 13.

16. Demuestre que si  $\beta(Q^k) > 0$  para algun  $k \geq 1$ . Entonces existe una única medida invariante  $\mu$  y

$$\sup_{a,y} |P(X_n^a = y) - \mu(y)| \leq (1 - \beta(Q^k))^{n/k}. \tag{14}$$

17. Determine if the chains presented in during the course are periodic and determine the period. For those matrices that are aperiodic and irreducible, determine the smallest power  $k$  satisfying that all the entries of  $Q^k$  are strictly positive.

18. Determine  $\beta(Q)$  and  $\alpha(Q)$  for all aperiodic and irreducible chains  $Q$  of Exercise 17. In case the computations become complicate, try to find bounds for  $\alpha(Q)$  and  $\beta(Q)$ . When  $\alpha(Q)$  gives a better convergence velocity than  $\beta(Q)$ ?

19. Let  $G = \{1, 2\}$  and  $Q$  be the following transition matrix

$$Q = \begin{pmatrix} \frac{1}{3} & \frac{2}{3} \\ \frac{2}{3} & \frac{1}{3} \end{pmatrix}$$

(a) Show that there exists  $\bar{n}$ , such that, for all  $n \geq \bar{n}$ ,

$$0,45 \leq Q^n(1, 2) \leq 0,55 \quad \text{and}$$

$$0,45 \leq Q^n(2, 2) \leq 0,55.$$

Find bounds for  $\bar{n}$ .

(b) Obtain similar results for  $Q^n(1, 1)$  and  $Q^n(2, 1)$ .