## Probabilistic Aspects of Computer Science: TD2 Markov chains in the long run

## Chargé de TD: Stéphane Le Roux

## 2022

**Exercise 1** (Move-to-front heuristic). Consider  $n \ge 1$  records  $R_1, R_2, \ldots, R_n$  that can be ordered in any order. The cost of accessing the *j*th record in the order is *j*. For example, with four records ordered as  $R_2, R_4, R_3, R_1$ , the cost of accessing  $R_4$  is 2 and the cost of accessing  $R_1$  is 4.

Assume that at every discrete time the record  $R_j$  is accessed with fixed positive probability  $p_j$ .

1. If we know the values of the  $p_j$ , how should we order the records to minimize the expected cost? What would be this cost?

In the remainder we assume that we do not know the  $p_j$ . To keep the cost low we use a *move-to-front* heuristic: at each step, put the record that was accessed at the front of the list (to the left). We assume that moving the record can be done at no cost and that the other records remain in the same order. For example, if the order is  $R_2, R_4, R_3, R_1$  before  $R_3$  is accessed, it becomes  $R_3, R_2, R_4, R_1$  afterwards.

- 2. Describe this problem by a Markov chain  $(X_k)_k$  whose state space is the set of words where the name or index of each register occurs exactly once. Describe  $P = M(p_1, \ldots, p_n)$  its transition probability matrix. Show that it admits a unique steady-state distribution.
- 3. The purpose of this question is to determine the steady-state distribution  $\pi$  of  $(X_k)_k$ . For a word w and a letter a, let  $w \setminus a$  be the result of removing all a's from w.
  - (a) Assume that  $n \ge 2$ . Let  $\sigma = \sigma_1 \dots \sigma_n$  be an ordering of the registers. Express  $\pi_{\sigma}$  using other orderings.
  - (b) Assume that  $n \ge 2$ . What is  $(X_k \setminus \sigma_1)_k$ ? Describe it qualitatively and quantitatively by a matrix Q.
  - (c) Find a relation between  $\pi_{\sigma}$  and  $\pi^Q_{\sigma\setminus\sigma_1}$ , where  $\pi^Q$  is the steady-state distribution of Q
  - (d) Assume that  $n \ge 2$ . Find a relation between Q and  $P' := M(\frac{p_{\sigma_2}}{1-p_{\sigma_1}} \dots \frac{p_{\sigma_n}}{1-p_{\sigma_1}})$ , and between their eigenvectors.
  - (e) Let

$$\varphi_n(x_1,\ldots,x_n) = \prod_{i=1}^n \frac{x_i}{\sum_{j=i}^n x_j}$$

Show that for all  $\sigma$  we have  $\pi_{\sigma} = \varphi_n(p_{\sigma_1}, \ldots, p_{\sigma_n})$  (the steady-state distribution of the chain).

4. The purpose of this question is to determine the expected cost of the move-to-front method. Let  $C_k$  be the cost for accessing the kth requested record,  $V_{k,i}$  be the event that the k-th query refers to record  $R_i$ , and  $W_{k,i}$  the Boolean random variable that equals 1 iff  $V_{k,i}$  holds. Using the law of total expectation we obtain the following equality.

$$\mathbf{E}(C_k) = \sum_{i=1}^{n} \mathbf{E}(C_k \mid V_{k,i}) \mathbf{Pr}(V_{k,i})$$

(a) For all k and  $i \neq j$  let  $L_{k,j,i} := 1$  if the register  $R_j$  is ordered before  $R_i$  just before the kth request, and  $L_{k,j,i} := 0$  otherwise. Under some assumption, find a relation between  $C_k$  and the  $L_{k,j,i}$ .

- (b) Are  $V_{k,i}$  and  $L_{k,j,i}$  dependent or independent?
- (c) Find a relation between  $\mathbf{E}(C_k \mid V_{k,i})$  and the  $L_{k,j,i}$ .
- (d) Describe the behavior of  $\mathbf{E}(L_{k,j,i})$  when k approaches  $+\infty$ .
- (e) Describe the behavior of  $\mathbf{E}(C_k)$  when k approaches  $+\infty$ , and compare with the first question.

Exercise 2. We revisit an exercise from last week using a new and more direct tool.

1. Let  $X_n$  be the number of heads obtained after n independent tosses of a (possibly unfair) coin. Show that, for any  $k \ge 2$ ,

$$\lim_{n \to \infty} \mathbf{Pr}(X_n \text{ is divisible by } k) = \frac{1}{k}$$

2. Solve the problem when  $X_n$  represents the sum of n independent rolls of a dice.

**Exercise 3.** Exhibit a Markov chain which has null recurrent states (different from the one studied during the lecture).

**Exercise 4.** Show that if a Markov chain has two steady-state distributions, then it has an infinite number of steady-state distributions.