Due: November 13 (Friday) at 23:59 (Pacific Time)

Please follow the homework policies on the course website.

## 1. (4 pt.) Coin Tosses Modulo k

A biased coin with probability of heads  $p \in (0,1)$  is tossed repeatedly. Let  $X_t$  denote the number of heads obtained in the first t tosses. Prove that for any integer  $k \geq 1$  and  $0 \leq j \leq k-1$ ,

$$\lim_{t \to +\infty} \Pr[(X_t \bmod k) = j] = \frac{1}{k}.$$

[HINT: If your proof applies the fundamental theorem of Markov chains at some point, make sure to prove that the Markov chain that you defined satisfies all the preconditions.]

## 2. (12 pt.) Fundamental Theorem of Markov Chains: A Special Case

Let  $X_0, X_1, \ldots$  be a Markov chain over n states (labeled  $1, 2, \ldots, n$ ) with transition matrix  $P \in \mathbb{R}^{n \times n}$ , i.e., for any  $t \geq 0$ ,  $\Pr[X_{t+1} = j | X_t = i] = P_{ij}$ . In addition, we assume that  $P_{ij} > 0$  for all  $i, j \in [n]$ , and define  $p_{\min} := \min_{i,j \in [n]} P_{ij} > 0$ . In this problem, we will prove part of the fundamental theorem of Markov chains for this special case. In particular, we will show that there exists a unique stationary distribution  $\pi$  such that for all  $i, j \in [n]$ ,

$$\lim_{t \to +\infty} \Pr[X_t = j | X_0 = i] = \pi_j.$$

- (a) (1 pt.) As a warmup, show that the assumption  $P_{ij} > 0$  for all  $i, j \in [n]$  implies that the Markov chain is irreducible and aperiodic. Thus, the assumption that we made is not weaker than the one in the original theorem.
- (b) **(4 pt.)** Suppose that vectors  $a, b \in \mathbb{R}^n$  satisfy  $\sum_{i=1}^n a_i = 0$  and  $\min_{i \in [n]} b_i \ge \epsilon > 0$ . Prove that  $|\sum_{i=1}^n a_i b_i| \le \sum_{i=1}^n |a_i| b_i \frac{\epsilon}{2} \sum_{i=1}^n |a_i|$ .

[HINT: It might be useful to define  $S^+ := \{i \in [n] : a_i \ge 0\}$  and  $S^- := \{i \in [n] : a_i < 0\}$ , and then relate  $\sum_{i=1}^n a_i b_i$  to  $\sum_{i \in S^+} |a_i| b_i$  and  $\sum_{i \in S^-} |a_i| b_i$ .

[HINT: It is possible to prove a stronger bound with the  $\frac{\epsilon}{2}$  factor replaced by  $\epsilon$ . ]

(c) **(2 pt.)** Let  $a = \begin{bmatrix} a_1 & a_2 & \cdots & a_n \end{bmatrix}$  be a row vector that satisfies  $\sum_{i=1}^n a_i = 0$ . Prove that  $||aP||_1 \le (1 - np_{\min}/2)||a||_1$ .

[HINT: Use the previous part. ]

(d) (3 pt.) Prove that there exists an *n*-dimensional row vector  $\pi = \begin{bmatrix} \pi_1 & \pi_2 & \cdots & \pi_n \end{bmatrix}$  such that: (1)  $\pi = \pi P$ ; (2)  $\sum_{i=1}^n \pi_i = 1$ .

**[HINT:** First prove the existence of a non-zero vector  $\pi$  satisfying  $\pi = \pi P$ , and then show that the second condition can be satisfied by scaling  $\pi$ . For the first step, you may use the following fact without proof: if  $\lambda$  is an eigenvalue of a square matrix A,  $\lambda$  is also an eigenvalue of  $A^T$ . Part 2c might be helpful for the second step.

(e) (2 pt.) Let  $v = \begin{bmatrix} v_1 & v_2 & \cdots & v_n \end{bmatrix}$  be a row vector that satisfies  $\sum_{i=1}^n v_i = 1$ . Let  $\pi$  be a vector chosen as in Part 2d. Prove that  $\lim_{t \to +\infty} vP^t = \pi$ . Then, derive that for all  $i, j \in [n]$ ,

$$\lim_{t \to +\infty} \Pr[X_t = j | X_0 = i] = \pi_j.$$

[HINT: Apply Part 2c to  $(v - \pi), (v - \pi)P, (v - \pi)P^2, \dots$ ]

(f) (0 pt.) [Optional: this won't be graded.] Extend the proof to the general case, where the Markov chain is irreducible and aperiodic but  $P_{ij} > 0$  might not hold.

## 3. (14 pt.) Coupon Collection on Markov Chains

Let  $X_1, X_2, \ldots$  be an irreducible and aperiodic Markov chain over  $n \geq 2$  states (labeled  $1, 2, \ldots, n$ ) with a uniformly distributed initial state and transition matrix  $P \in \mathbb{R}^{n \times n}$ , i.e.,  $\Pr[X_1 = i] = 1/n$  for all  $i \in [n]$  and  $\Pr[X_{t+1} = j | X_t = i] = P_{ij}$  for any  $t \geq 1$  and  $i, j \in [n]$ . Furthermore, we assume that P is doubly stochastic, which means that  $\sum_{i=1}^{n} P_{ij} = 1$  for every  $j \in [n]$ , as well as  $\sum_{j=1}^{n} P_{ij} = 1$  for  $i \in [n]$ . (Note that a general transition matrix need only satisfy the second of these).

- (a) (1 pt.) Prove that the unique stationary distribution of this Markov chain is uniform over [n]. Moreover, show that  $\Pr[X_t = i] = 1/n$  for every  $t \ge 1$  and  $i \in [n]$ .
- (b) (5 pt.) Let T denote the earliest time when all the states have been visited at least once and the chain returns to the initial state  $X_1$ . Formally, random variable T is defined as

$$T = \min\{t \ge 1 : X_t = X_1 \text{ and } \{X_1, X_2, \dots, X_t\} = [n].\}$$

Consider the following argument for bounding  $\mathbb{E}[T]$ :

The conclusion of Part 3a says that each  $X_t$  is uniformly distributed over [n]. Therefore,  $X_1, X_2, \ldots$  can be viewed as the coupons that we draw in the coupon collector's problem. Then, T is exactly the number of coupons needed for first collecting the n coupons and then waiting for coupon  $X_1$  to appear again. In expectation, the first part takes  $n \sum_{i=1}^{n} \frac{1}{i}$  steps and the second part takes  $\frac{1}{1/n} = n$  steps, so  $\mathbb{E}[T] = n \sum_{i=1}^{n} \frac{1}{i} + n = \Theta(n \log n)$ .

Disprove the above by doing the following: (1) point out the flaw in the reasoning; (2) construct counterexamples to show that  $\mathbb{E}[T]$  might be as small as O(n); (3) construct counterexamples to show that  $\mathbb{E}[T]$  might not be upper bounded by any function of n. More concretely, for Part (2), show that for every  $n \geq 2$  there exists a Markov chain with n states satisfying all the assumptions and  $\mathbb{E}[T] \leq 2n$ . (You will also get full credit if  $\mathbb{E}[T] \leq Cn$  for some other constant C.)

For Part (3), show that for every M > 0 there exists a Markov chain with n = 2 states satisfying all the assumptions and  $\mathbb{E}[T] \geq M$ . (You will also get full credit if the number of states is n = C for some other constant C.)

- (c) (3 pt.) Define  $p_{\min} := \min_{i,j \in [n]} P_{ij}$ . Prove that if  $p_{\min} > 0$ , we have  $\mathbb{E}[T] \leq O\left(\frac{\log n}{p_{\min}}\right)$ . [HINT: It can be proved that  $\mathbb{E}[T] \leq 1 + \frac{1}{p_{\min}}\left(1 + \sum_{i=1}^{n-1} \frac{1}{i}\right)$ .]
- (d) (5 pt.) Let  $\mu := \mathbb{E}[T]$ . Show that T has a *sub-exponential tail*, i.e., that there exist constants  $c_0, c_1 > 0$  (that do not depend on n or any other parameters of the Markov chain) such that for every  $\lambda \geq c_0$ :

$$\Pr[T > \lambda \mu] \le e^{-c_1 \lambda}.$$

<sup>&</sup>lt;sup>1</sup>Note that in this problem, the initial state is denoted by  $X_1$  (instead of  $X_0$ , which is used in the lecture notes). This slight change makes the analogy to coupon collecting in Part 3b more natural.

You may use the following claim without proof:

Claim 1. There exists a constant  $c_2 > 0$  such that the following holds for any Markov chain with n states that satisfies all the assumptions, and for any  $i \in [n]$ : Let  $M = \lfloor c_2 \mu \rfloor$ . Conditioning on  $X_1 = i$ , with probability at least 1 - 1/e, there exists  $t \in [M]$  such that  $\{X_1, X_2, \ldots, X_t\} = \{X_{t+1}, X_{t+2}, \ldots, X_M\} = [n]$ .

[HINT: You might find the argument at the end of the proof of Theorem 1 in Lecture Note #13 useful.]

(e) (0 pt.) [Optional: this won't be graded.] Prove Claim 1.