

MathTools HW 8

1. **The ergodic theorem for Markov chains.** Consider an ergodic Markov chain over a finite state space \mathcal{S} , with the stationary distribution π . Recall that for any $f : \mathcal{S} \rightarrow \mathbb{R}$, the averages $\frac{1}{T} \sum_{t=1}^T f(X_t)$ converge in probability,¹ as $T \rightarrow \infty$, to the expectation $\mathbb{E}_{X \sim \pi}[f(X)]$.

- (a) For a state $s \in \mathcal{S}$, let T_s be the amount of time the chain spends in s along the first T time steps, in other words,

$$T_s = |\{1 \leq t \leq T : X_t = s\}|.$$

T_s is, of course, a random variable. Show that² the proportion of time spent in s , $T_s/T \xrightarrow{T \rightarrow \infty} \pi_s$.

- (b) Extend the ergodic theorem, in the following manner. Prove that for any $f : \mathcal{S} \times \mathcal{S} \rightarrow \mathbb{R}$,

$$\frac{1}{T} \sum_{t=1}^T f(X_t, X_{t+1}) \xrightarrow{T \rightarrow \infty} \mathbb{E}[f(Y_1, Y_2)],$$

where Y_1, Y_2 are random variables that have the joint distribution $\Pr(Y_1 = a, Y_2 = b) = \pi_a P_{ab}$ (that is, first sample Y_1 according to π , and then sample Y_2 according to the conditional distribution of the chain).

Hint: Construct an ergodic Markov chain whose state space, \mathcal{T} , is a subset of $\mathcal{S} \times \mathcal{S}$. Apply the ergodic theorem for that chain.

- (c) For $a, b \in \mathcal{S}$, let $T_{a,b}$ be the amount of time the chain spends on the edge (a, b) , that is,

$$T_{a,b} = |\{1 \leq t \leq T : X_t = a \text{ and } X_{t+1} = b\}|.$$

Show that $T_{a,b}/T \xrightarrow{T \rightarrow \infty} \pi_a P_{ab}$.

2. **Gambler's ruin.** Consider the following 2-player game. Player 1 starts with k coins, and player 2 starts with $n - k$ coins. At each round, a fair coin is flipped; if the coin lands head, player 1 gains a coin and player 2 loses one; otherwise, player 1 loses a coin and player 2 gains one. The game ends once one of the players loses all his coins (and the other gains all n coins).

- (a) Let X_t be the number of coins of player 1 at round t . Explain why the process X_1, X_2, \dots constitutes a Markov chain (describe its transition matrix). Show that this chain is *not* irreducible.

Henceforth, denote the transition matrix of the chain by P .

- (b) Let \mathbf{q} be any starting distribution. Prove that

$$\lim_{t \rightarrow \infty} (\mathbf{q}^\top P^t)_k = 0 \quad \text{for all } 1 \leq k \leq n - 1.$$

¹We say that a sequence of random variables, Z_1, Z_2, \dots converges in probability to a number $a \in \mathbb{R}$ if for every $\epsilon > 0$, the probability $\Pr(|Z_n - a| \geq \epsilon) \rightarrow 0$ as $n \rightarrow \infty$.

²Convergence here = convergence in probability.

- (c) Denote by \mathbf{e}_k the probability that puts all its mass on k . Show that the stationary distributions of P are exactly the convex combinations of \mathbf{e}_0 and \mathbf{e}_b , that is, all the distributions of the form

$$\boldsymbol{\pi} = \alpha \mathbf{e}_0 + (1 - \alpha) \mathbf{e}_b, \quad \text{where } \alpha \in [0, 1].$$

Remark: $\boldsymbol{\pi}$ is a stationary distribution of P whenever $\boldsymbol{\pi}^\top P = \boldsymbol{\pi}^\top$.

- (d) Compute $\lim_{t \rightarrow \infty} \mathbf{e}_k^\top P^t$ for all k .

Hint: Let p_k be the probability that, assuming player 1 starts with k coins, he eventually loses them all. In other words, it is the probability, given that the initial state is $X_1 = k$, that the chain eventually hits the state 0. Find a recursive formula for p_k ; guess a solution (or solve the equation systematically) - it should be very easy.

3. **Expander mixing lemma.** Let G be a d -regular graph on n vertices. Let A_G be the adjacency matrix of G , and denote its eigenvalues by $\lambda_1 \geq \dots \geq \lambda_n$. Recall that $\lambda_1 = d$, and denote $\lambda^* = \max\{|\lambda_2|, |\lambda_n|\}$. For sets $S, T \subset [n]$, denote by $e(S, T)$ the number of edges the cross between S and T , where edges that start and end in $S \cap T$ are counted twice; that is,

$$e(S, T) = |\{(s, t) : s \in S, t \in T\}|$$

(the number of *ordered* pairs). Prove that

$$\left| e(S, T) - \frac{d}{n} |S| |T| \right| \leq \lambda^* \sqrt{|S| |T|}.$$

Remark: Note that in a random graph $\mathcal{G}(n, d/n)$, with average degree d , one has $\mathbb{E}[e(S, T)] = \frac{d}{n} |S| |T|$ (say when S, T are disjoint). This result says that expanders (in the sense of small λ^*) behave, in some sense, like random graphs.

4. **Max-Cut.** Let G be a graph on n vertices. A *cut* is a partition of its vertices in two parts (S, S^c) (where $S^c = [n] \setminus S$). The *size* of the cut is $e(S, S^c)$, the number of edges that cross the cut (see previous question). The maximum cut problem (Max-Cut) asks for the size of a maximum cut in G , denoted here by

$$c(G) = \max_{S \subset [n]} e(S, S^c).$$

As many of you probably know, the maximum cut problem is in general NP-complete.

Assume that G is d -regular. Prove the following bounds on the maximum cut of G :

- (a) $c(G) \leq \frac{n(d - \lambda_n)}{4}$, where λ_n is the *smallest* eigenvalue of the adjacency matrix.

Hint: The bound of Q(3) is unfortunately a bit too loose to prove this. Letting $\mathbf{1}_S$ be the indicator for the set S , analyze $\mathbf{1}_S^\top A_G \mathbf{1}_{S^c}$, where also note that $\mathbf{1}_{S^c} = \mathbf{1} - \mathbf{1}_S$.

- (b) $c(G) \geq \frac{nd}{4}$.

Hint: Use the *probabilistic method*. Choose S randomly, and compute $\mathbb{E}[e(S, S^c)]$.