

# AI 5090: STOCHASTIC PROCESSES

## HOMEWORK 6



### TOPICS: MARKOV CHAINS, SAMPLING TECHNIQUES

Fix a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ .

Assume that all random variables appearing below are defined with respect to this probability space.

1. Let  $\{X_n\}_{n=0}^\infty$  be an irreducible, time-homogeneous DTMC on a finite state space  $\mathcal{X}$ , with TPM  $P$  and unique stationary distribution  $\pi$ . Given two disjoint sets  $A, B \subseteq \mathcal{X}$ ,  $A \cap B = \emptyset$ , define the **probability flux** from  $A$  to  $B$  as

$$\Phi(A, B) := \sum_{x \in A} \sum_{y \in B} \pi_x P_{x,y}.$$

- (a) Let  $\mathcal{X}_1 \subset \mathcal{X}$  and  $\mathcal{X}_2 \subset \mathcal{X}$  form a partition of  $\mathcal{X}$ . Show that probability flux balances, i.e.,

$$\Phi(\mathcal{X}_1, \mathcal{X}_2) = \Phi(\mathcal{X}_2, \mathcal{X}_1).$$

- (b) Rewrite the relation  $\pi = \pi P$  as a flux balance equation.

2. Let  $\{X_n\}_{n=0}^\infty$  be a time-homogeneous, **ergodic** Markov chain on a discrete state space  $\mathcal{X}$ , with TPM  $P$  and unique stationary distribution  $\pi$ .

Let  $\{Y_n\}_{n=0}^\infty$  be an independent copy of  $\{X_n\}_{n=0}^\infty$  with the same TPM  $P$  and  $Y_0 \sim \pi$ .

Let  $\tau = \inf\{n \geq 0 : X_n = Y_n\}$  be the coupling time.

In class, we showed that if  $X_0 = x$ , then

$$|P_{x,y}^n - \pi(y)| \leq 2\mathbb{P}(\tau > n \mid X_0 = x) \quad \forall y \in \mathcal{X}.$$

- (a) Show that there exists  $\lambda \in (0, 1)$  such that

$$\mathbb{P}(\tau > n \mid X_0 = x) \leq \lambda^n \quad \forall n \geq 0.$$

**Hint:** Note that

$$\mathbb{P}(\tau > n \mid X_0 = x) = \mathbb{P}(X_0 \neq Y_0, \dots, X_n \neq Y_n \mid X_0 = x).$$

Express the probability on the right-hand side above using the entries of  $P$  and  $\pi$ .

What do you know about the entries of  $\pi$ ?

- (b) Let  $N_y(n) = \sum_{k=1}^n \mathbf{1}_{\{X_k=y\}}$  denote the number of visits to state  $y$  up to time  $n$ . Using the result of part (a), prove that starting from state  $x$ ,

$$\frac{N_y(n)}{n} \xrightarrow{\text{m.s.}} \pi_y.$$

- (c) Using the result of part (a) and the Borel–Cantelli lemma, prove that starting from state  $x$ ,

$$\frac{N_y(n)}{n} \xrightarrow{\text{a.s.}} \pi_y.$$

3. Let  $\{X_n\}_{n=0}^\infty$  be a time-homogeneous DTMC on a discrete state space  $\mathcal{X}$  and TPM  $P$ . Assume that  $P_{x,x} < 1$  for all  $x \in \mathcal{X}$ . Let

$$T_1 := \inf\{n \in \mathbb{N} : X_n \neq X_0\},$$

and for  $m \geq 2$ , let

$$T_m := \inf\{n \in \mathbb{N} : X_n \neq X_{T_{m-1}}\}.$$

Thus,  $T_m$ 's denote the random times at which the DTMC changes state.

- (a) For each  $m \in \mathbb{N}$ , show that  $T_m$  is a stopping time w.r.t. the natural filtration of the process  $\{X_n\}_{n=0}^\infty$ .

- (b) Let  $Z_0 = X_0$ , and for each  $m \in \mathbb{N}$ , let  $Z_m = X_{T_m}$ . Prove that  $\{Z_m\}_{m=0}^\infty$  is a time-homogeneous DTMC with TPM  $\tilde{P}$  given by

$$\tilde{P}_{x,y} = \begin{cases} 0, & x = y, \\ \frac{P_{x,y}}{1 - P_{x,x}}, & x \neq y. \end{cases}$$

- (c) If  $P$  admits a unique stationary distribution  $\pi$ , determine the stationary distribution of  $\tilde{P}$ .
4. Fix a sufficiently large number  $N \in \mathbb{N}$ , and let  $\{X_n\}_{n=0}^N$  be an irreducible and positive recurrent DTMC with TPM  $P$  and stationary distribution  $\pi$ . Suppose that  $X_0 \sim \pi$ . For each  $n \in \{0, \dots, N\}$ , define  $Y_n = X_{N-n}$ . The process  $\{Y_n\}_{n=0}^N$  is called the time-reversed chain.

- (a) From an earlier homework, we know that  $\{Y_n\}_{n=0}^N$  is a DTMC. Identify its TPM.
- (b) What conditions should the entries of  $P$  and  $\pi$  satisfy for  $\{X_n\}_{n=0}^N$  and  $\{Y_n\}_{n=0}^N$  to have identical TPMs?

**Remark:** A DTMC  $\{X_n\}_{n=0}^N$  which starts off in its stationary distribution is said to be **reversible** chain if its TPM is identical to the TPM of its time-reversed version.

5. We saw the rejection sampling technique in class. Prove formally that

$$\mathbb{P}(Z \leq x \mid E) = F(x),$$

where  $E = \{a U f_Z(Z) \leq f(Z)\}$ , the function  $f$  is the PDF corresponding to the CDF  $F$ , and  $(U, Z)$  satisfy the conditions outlined in the lecture with

$$f(x) \leq a f_Z(x) \quad \forall x.$$

6. **Another variant of rejection sampling.**

Fix a target PDF  $f$  with range  $[0, \infty)$ . Let  $U_1, U_2 \stackrel{\text{i.i.d.}}{\sim} \text{Unif}(0, 1)$ . Define  $R = U_2/U_1$ , and let  $E$  denote the event

$$E = \left\{ U_1 \leq \sqrt{f(U_2/U_1)} \right\}.$$

Show that

$$\mathbb{P}(R \leq x \mid E) = \int_0^x f(t) dt.$$

**Hint:** Use the transformation  $S = U_2/U_1, T = U_1$ .

The idea is that we sample two independent uniform  $(0, 1)$  random variables, and check for the condition in event  $E$ . If this condition is satisfied, we accept the uniform samples and simply output  $R$  as the desired sample. If not, we reject the uniform samples and repeat the procedure till the condition in  $E$  holds. This variant of rejection sampling technique holds only when the target sample desired has non-negative support.