

"Markov Processes", Problem Sheet 9.

Hand in solutions before Friday 12.12., 2 pm (post-box opposite to maths library)

1. (Infinitesimal characterization of stationary distributions) Consider a timehomogeneous continuous time Markov chain $X_t = Y_{N_t}$ where (N_t) is a Poisson process with constant intensity $\lambda > 0$, and (Y_n) is an independent Markov chain with transition matrix π on a finite state space S.

a) Show that the transition function is given by

$$p_t(x,y) = P_x[X_t = y] = \exp(t\mathcal{L})(x,y),$$

where $\mathcal{L} = \lambda(\pi - I)$ and $\exp(t\mathcal{L})$ is the matrix exponential. Hence conclude that $(p_t)_{t\geq 0}$ satisfies the forward and backward equation

$$\frac{d}{dt}p_t = p_t \mathcal{L} = \mathcal{L}p_t \quad \text{for } t \ge 0.$$

b) Prove that a probability measure μ on S is a stationary distribution if and only if

$$\sum_{x \in S} \mu(x) \mathcal{L}(x, y) = 0 \quad \text{for any } y \in S.$$

c) Show that the transition matrices are self-adjoint in $L^2(\mu)$, i.e.,

$$\sum_{x \in S} f(x) (p_t g)(x) \mu(x) = \sum_{x \in S} (p_t f)(x) g(x) \mu(x) \quad \text{for any } t \ge 0, \ f, g : S \to \mathbb{R},$$

if and only if the generator \mathcal{L} satisfies the detailed balance condition w.r.t. μ . What does this mean for the process ?

2. (Simple exclusion process) Let $\mathbb{Z}_n^d = \mathbb{Z}^d/(n\mathbb{Z})^d$ denote a discrete *d*-dimensional torus. The simple exclusion process on $S = \{0, 1\}^{\mathbb{Z}_n^d}$ is the Markov process with generator

$$(\mathcal{L}f)(\eta) = \frac{1}{2d} \sum_{x \in \mathbb{Z}_n^d} \sum_{y: |y-x|=1} 1_{\{\eta(x)=1, \eta(y)=0\}} \cdot (f(\eta^{x,y}) - f(\eta)),$$

where $\eta^{x,y}$ is the configuration obtained from η by exchanging the values at x and y. Show that any Bernoulli measure of type

$$\mu_p = \bigotimes_{x \in \mathbb{Z}_n^d} \nu_p, \qquad \nu_p(1) = p, \ \nu_p(0) = 1 - p,$$

 $p \in [0, 1]$, is a stationary distribution. Why does this not contradict the fact that any irreducible Markov process on a finite state space has a unique stationary distribution? (You may assume the statements of Exercise 1).

3. (Bounds for ergodic averages in the non-stationary case) Let $(X_n)_{n \in \mathbb{Z}_+}$ be a Markov chain on (S, \mathcal{B}) with transition kernel p and stationary distribution μ , and let

$$A_{b,n}f = \frac{1}{n} \sum_{i=b}^{b+n-1} f(X_i).$$

Assume that there are a distance d on S, $0 < \alpha < 1$ and $\bar{\sigma} \in \mathbb{R}_+$ such that

(A1) $\mathcal{W}_d^1(\nu p, \tilde{\nu} p) \leq \alpha \mathcal{W}_d^1(\nu, \tilde{\nu}) \quad \forall \nu, \tilde{\nu} \in \mathcal{P}(S), \text{ and}$

(A2) $\operatorname{Var}_{p(x,\cdot)}(f) \leq \bar{\sigma}^2 ||f||^2_{Lip(d)} \quad \forall x \in S, f : S \to \mathbb{R}$ Lipschitz.

Prove that under these assumptions the following bounds hold for any $b, n, k \ge 0, x \in S$, and for any Lipschitz continuous function $f: S \to \mathbb{R}$:

- a) $\operatorname{Var}_{x}[f(X_{n})] \leq \sum_{k=0}^{n-1} \alpha^{2k} \bar{\sigma}^{2} \|f\|_{Lip(d)}^{2}$.
- b) $|\operatorname{Cov}_x[f(X_n), f(X_{n+k})]| = |\operatorname{Cov}_x[f(X_n), (p^k f)(X_n)]| \le \frac{\alpha^k}{1-\alpha^2}\bar{\sigma}^2||f||^2_{Lip(d)}.$

c)
$$\operatorname{Var}_{x}[A_{b,n}f] \leq \frac{1}{n} \frac{\bar{\sigma}^{2}}{(1-\alpha)^{2}} \|f\|_{Lip(d)}^{2}$$
.

d)
$$|E_x[A_{b,n}f] - \mu(f)| \leq \frac{1}{n} \frac{\alpha^b}{1-\alpha} \int d(x,y) \, \mu(dy) \, ||f||_{Lip(d)}.$$

e) $E_x \left[|A_{b,n}f - \mu(f)|^2 \right] \leq \frac{1}{n} \frac{1}{(1-\alpha)^2} \left(\bar{\sigma}^2 + \frac{1}{n} \alpha^{2b} (\int d(x,y) \, \mu(dy))^2 \right) \|f\|_{Lip(d)}^2$.

4. (Succesful couplings and TV-convergence to equilibrium) Consider a Markov chain on (S, \mathcal{B}) with transition kernel p and stationary distribution μ . A coupling (X_n, Y_n) of the chains with initial distributions ν and μ respectively is called *succesful* if the coupling time

$$T = \inf \{ n \ge 0 : X_n = Y_n \text{ for any } n \ge T \}$$

is almost surely finite. Show that a successful coupling exists if and only if $||\nu p^n - \mu||_{TV} \to 0$ as $n \uparrow \infty$.