

Example Sheet 2

1. Let  $X$  be a reversible Markov chain on a finite state space  $E$  with transition matrix  $P$  and invariant distribution  $\pi$ . Prove a generalisation of the Poincaré inequality, i.e. for all  $f : E \rightarrow \mathbb{R}$  show that

$$\text{Var}_\pi(P^t f) \leq e^{-2t/t_{\text{rel}}} \text{Var}_\pi(f).$$

2. Let  $P$  be a reversible transition matrix on a finite state space with invariant distribution  $\pi$ . Define the total variation distance from stationarity from a typical point, i.e. for all  $t$

$$d_{\text{ave}}(t) = \sum_x \pi(x) \|P^t(x, \cdot) - \pi\|_{\text{TV}}.$$

Suppose that  $1 = \lambda_1 \geq \dots \geq \lambda_n \geq -1$  are the eigenvalues, then show that

$$4d_{\text{ave}}(t)^2 \leq \sum_{j=2}^n \lambda_j^{2t}.$$

3. Consider a lazy simple random walk on  $\mathbb{Z}_n$ . Show that for all  $\alpha > 0$  we have as  $n \rightarrow \infty$

$$d(\alpha n^2) \rightarrow \int_0^1 \left| \sum_{k=1}^{\infty} e^{-\alpha \pi^2 k^2} \cos(2\pi k u) \right| du.$$

(Hint: First write

$$d(\alpha n^2) = \frac{1}{2} \int_0^1 \left| 1 - n P^{\lceil \alpha n^2 \rceil}(0, \lfloor un \rfloor) \right| du$$

and then use the spectral theorem.)

4. Let  $X$  be an irreducible Markov chain on the finite state space  $E$  with transition matrix  $P$  and invariant distribution  $\pi$ .

(i) Define the separation distance  $s(t) = \max_{x,y} (1 - P^t(x,y)/\pi(y))$ . Show that  $s(t)$  is decreasing as a function of  $t$ .

(ii) Define  $t_{\text{sep}}(\varepsilon) = \min\{t \geq 0 : s(t) \leq \varepsilon\}$ . Show that for all  $\varepsilon \in (0, 1]$  and all  $k \in \mathbb{N}$  we have that

$$t_{\text{sep}}(\varepsilon^k) \leq k t_{\text{sep}}(\varepsilon).$$

5. Consider two copies  $K_n$  and  $K'_n$  of the complete graph joined by a single edge. Find the order of the mixing time for a lazy simple random walk on the resulting graph.

6. Let  $X$  be an irreducible, lazy and reversible Markov chain on a finite state space with transition matrix  $P$  and stationary distribution  $\pi$ . Let  $1 = \lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$  be the eigenvalues of  $P$ .

(i) Show that

$$\mathbb{E}_\pi[\tau_\pi] := \sum_{x,y} \pi(x)\pi(y)\mathbb{E}_x[\tau_y] = \sum_{i \geq 2} \frac{1}{1 - \lambda_i}.$$

(Hint: Use question 12(b) from the first example sheet.)

(ii) Show that

$$\sum_{t=k}^{\infty} (P^t(x, x) - \pi(x)) \leq e^{-k/t_{\text{rel}}} \mathbb{E}_{\pi}[\tau_x].$$

**7.** Let  $X$  be a reversible Markov chain on the finite state space  $E$  with transition matrix  $P$  and invariant distribution  $\pi$ .

(i) Prove that for all  $x, y$

$$\frac{P^{2t}(x, y)}{\pi(y)} \geq \left(1 - \max_{z, w} \|P^t(z, \cdot) - P^t(w, \cdot)\|_{\text{TV}}\right)^2.$$

Deduce that

$$P^{2t_{\text{mix}}}(x, y) \geq \frac{1}{4}\pi(y)$$

and that there exists a transition matrix  $\tilde{P}$  such that

$$P^{2t_{\text{mix}}}(x, y) = \frac{1}{4}\pi(y) + \frac{3}{4}\tilde{P}(x, y)$$

(ii) Define

$$t_{\text{stop}} = \max_x \min\{\mathbb{E}_x[\Lambda_x] : \Lambda_x \text{ is a randomised stopping time s.t. } \mathbb{P}_x(X_{\Lambda_x} \in \cdot) = \pi(\cdot)\}.$$

(It is not clear by the definition that a stationary time achieving the minimum exists. One such example is the filling rule introduced by Baxter and Chacon.) By defining an appropriate stationary time, prove that

$$t_{\text{stop}} \leq 8t_{\text{mix}}.$$

We say that a randomised stopping time  $T$  starting from  $x$  has a halting state if there exists  $z \in E$  such that  $T \leq \tau_z$ , where  $\tau_z = \min\{t \geq 0 : X_t = z\}$ .

(Harder) Show that if  $T$  has a halting state, then it is mean optimal, in the sense that

$$\mathbb{E}_x[T] = \min\{\mathbb{E}_x[\Lambda_x] : \Lambda_x \text{ is a randomised stopping time s.t. } \mathbb{P}_x(X_{\Lambda_x} \in \cdot) = \pi(\cdot)\}.$$

(Hint: For a stopping time  $S$  consider the exit frequencies from each state, i.e.  $\nu(y) = \mathbb{E}_x\left[\sum_{k=0}^{S-1} \mathbf{1}(X_k = y)\right]$  for all  $y$  and compare them for different stopping times. Then use the uniqueness of the invariant measure up to multiplying by a constant.)

**8.** Let  $X$  be a reversible Markov chain with values in the finite space  $E$ , transition matrix  $P$  and invariant distribution  $\pi$ .

(a) Let  $\varphi$  be an eigenfunction of  $P$  corresponding to eigenvalue  $\lambda \neq 1$  and  $\|\varphi\|_2 = 1$ . Show that

$$\mathbb{E}_{\pi} \left[ \left( \sum_{s=0}^{t-1} \varphi(X_s) \right)^2 \right] \leq \frac{2t}{1-\lambda}.$$

(b) Let  $f : E \rightarrow \mathbb{R}$  be a function with  $\mathbb{E}_{\pi}[f] = 0$ . Recall  $\gamma = 1 - \lambda_2$  is the spectral gap. Show that

$$\mathbb{E}_{\pi} \left[ \left( \sum_{s=0}^{t-1} f(X_s) \right)^2 \right] \leq \frac{2t\mathbb{E}_{\pi}[f^2]}{\gamma}.$$

(c) Using coupling or otherwise, show that if  $r \geq t_{\text{mix}}(\varepsilon/2)$  and  $t \geq 4t_{\text{rel}}\text{Var}_\pi(f)/(\eta^2\varepsilon)$ , then for all  $x \in E$

$$\mathbb{P}_x \left( \left| \frac{1}{t} \sum_{s=0}^{t-1} f(X_{r+s}) - \mathbb{E}_\pi[f] \right| \geq \eta \right) \leq \varepsilon.$$

**9.** Let  $X$  be a reversible Markov chain on a finite state space  $E$  with transition matrix  $P$  and invariant distribution  $\pi$ . Let  $A \subsetneq E$  and let  $B = A^c$  with  $k = |B|$ . Suppose that the sub-stochastic matrix  $P_B$  (the restriction of  $P$  to  $B$ , i.e.  $P_B(x, y) = P(x, y)$  for  $x, y \in B$ ) is irreducible, in the sense that for all  $x, y \in B$ , there exists  $n \geq 0$  such that  $P_B^n(x, y) > 0$ .

(i) By defining an appropriate inner product, show that  $P_B$  has  $k$  real eigenvalues

$$1 \geq \gamma_1 > \gamma_2 \geq \dots \geq \gamma_k.$$

(ii) Show that there exist nonnegative numbers  $a_1, \dots, a_k$  satisfying  $\sum_i a_i = 1$  such that for all  $t \geq 0$  we have

$$\mathbb{P}_{\pi_B}(\tau_A > t) = \sum_{i=1}^k a_i \gamma_i^t,$$

where  $\pi_B(x) = \pi(x)/\pi(B)$  for all  $x \in B$ .

(iii) The Perron Frobenius theorem gives that  $\gamma_1 > 0$  and  $\gamma_1 \geq -\gamma_k$ . Using the Courant-Fischer characterisation of eigenvalues establish that

$$\gamma_1 \leq 1 - \frac{\pi(A)}{t_{\text{rel}}}.$$

(iv) Deduce that  $\mathbb{P}_{\pi_B}(\tau_A > t) \leq \left(1 - \frac{\pi(A)}{t_{\text{rel}}}\right)^t \leq \exp\left(-\frac{t\pi(A)}{t_{\text{rel}}}\right)$ .

(v) By the Perron Frobenius theorem the left eigenvector  $v$  corresponding to  $\gamma_1 > 0$  is strictly positive. Let  $\alpha$  be a probability distribution given by  $\alpha = v/\sum_i v(i)$ . Show that when the starting distribution is  $\alpha$ , then the law of  $\tau_A$  is geometric with parameter  $\gamma_1$ .

Prove that for all  $t$  and all  $y$

$$\mathbb{P}_\alpha(X_t = y \mid \tau_A > t) = \alpha(y).$$

Finally show that if  $P_B$  is in addition aperiodic, then for all  $x \notin A$  we have

$$\mathbb{P}_x(X_t = y \mid \tau_A > t) \rightarrow \alpha(y) \text{ as } t \rightarrow \infty.$$

(The distribution  $\alpha$  is called the quasi-stationary distribution.)