

Submit solutions to 3 problems among problems 1 and 4-10. You are allowed to discuss the problems with your colleagues, but you have to write your own solution, and specify in the solution who did you solve/discuss the problem with. Due: Feb 11th.

Example Sheet 1

1. (a) Suppose  $X$  is a random walk on a graph of diameter  $D$  (that is, the maximal distance between vertices of the graph). Show that  $t_{\text{mix}}(\delta) \geq D/2$  for all  $\delta < 1/2$ .  
 (b) Argue that also the lazy random walk satisfies the above, and that the rate 1 continuous-time random walk satisfies  $t_{\text{mix}} \gtrsim D$ . (No need to submit solution for this part. Just check you understand why this is the case.)  
 (c) We say that a sequence of graphs  $G_n = (V_n, E_n)$  is an *expander family* if  $\gamma^{(n)} \asymp 1$ , where  $\gamma^{(n)}$  is the spectral-gap of the transition matrix of SRW on  $G_n$ . Deduce that if  $G_n$  is a sequence of  $d$ -regular graphs for a fixed  $d$  (i.e.  $\deg(x) = d$  for all  $x \in V_n$  for all  $n$ ) and is an expander family then the lazy SRW and the rate 1 continuous-time SRW satisfy that for all fixed  $\delta \in (0, 1)$  and all  $\varepsilon > 0$  we have for  $n$  sufficiently large that

$$(1 - \varepsilon) \log_d(\delta n) \leq t_{\text{mix}}(1 - \delta) \leq t_{\text{mix}}^{(2)}(1/n^2) \lesssim \log n.$$

Deduce that there is pre-cutoff.

2. Let  $X, X'$  be  $\text{Poisson}(\lambda, \lambda')$  respectively, and let  $\mu, \mu'$  be their respective law. By a coupling argument or otherwise, show that  $\|\mu - \mu'\|_{TV} \leq |\lambda' - \lambda|$  for all  $\lambda, \lambda'$ .
3. Consider the coupon collector problem: an urn contains  $n$  white balls initially and we sample from the urn with replacement uniformly at random, each time painting the ball black (whatever its colour). Show that the time  $\tau_n$  until all balls are black is concentrated near  $n \log n$ , i.e.,  $\tau_n/(n \log n) \rightarrow 1$  in probability. How many balls are black by time  $(1/2)n \log n$ ?
4. (a) Let  $\mathbb{P}$  be the  $\text{Binomial}(n, 1/2)$  distribution and let  $\mathbb{Q}$  and  $\mu$  be the  $\text{Binomial}(n - m_+, \frac{1}{2})$  and  $\text{Binomial}(n - m_-, \frac{1}{2})$  distributions (respectively), where  $m_+ = o(n^{1/2})$  and  $n^{1/2} = o(m_-)$ . Show that as  $n \rightarrow \infty$  we have that

$$\|\mathbb{P} - \mathbb{Q}\|_{TV} \rightarrow 0 \quad \text{and} \quad \|\mathbb{P} - \mu\|_{TV} \rightarrow 1.$$

- (b) Now consider lazy random walk  $(Z_t : t = 0, 1, \dots)$  on the hypercube  $\{0, 1\}^n$  started from  $(0, \dots, 0)$  (the mixing time is independent on the initial state - make sure you understand why this is the case!): at each step we select a coordinate uniformly at random, and flip it with probability  $1/2$ . Show that at equilibrium, the number of coordinates equal to 1 follows the  $\text{Binomial}(n, 1/2)$  distribution. Let  $\tau_{\pm}$  be the time at which all but  $m_{\pm} = n^{1/2}(\log n)^{\mp 1}$  coordinates have been selected. Show that  $\mathbb{P}(|\frac{2\tau_{\pm}}{n \log n} - 1| \geq \epsilon) \rightarrow 0$  as  $n \rightarrow \infty$ , for any  $\epsilon > 0$ .  
 (c) Let  $\pi$  be the equilibrium distribution. Show that  $\tau_+$  is independent of  $Z_{\tau_+}$  and that

$$\|\mathbb{P}(Z_{\tau_+} = \cdot) - \pi\|_{TV} = \|\text{Binomial}(n, 1/2) - \text{Binomial}(n - m_+, 1/2)\|_{TV} \rightarrow 0$$

and that

$$\begin{aligned} \|\mathbb{P}(Z_k = \cdot) - \pi\|_{TV} &\geq \mathbb{P}[\tau_- > k] \min_{i \in (m_-, n]} \|\text{Binomial}(n, 1/2) - \text{Binomial}(n - i, 1/2)\|_{TV} - o(1) \\ &= (1 - o(1))\mathbb{P}[\tau_- > k] - o(1). \end{aligned}$$

(d) Deduce that  $Z$  exhibits cutoff at time  $(1/2)n \log n$ . (Make sure you understand why the independence of  $Z_{\tau_+}$  from  $\tau_+$  is crucial for bounding  $d(t)$  in terms of  $\mathbb{P}(\tau_+ > t)$ .)

5. Random transpositions is the Markov chain such that at each step the left hand picks a card uniformly in the deck, the right hand too, and the two cards are switched. (Note that the two cards can be the same, in which case nothing happens). Formally,

$$P(\sigma, \sigma') = \begin{cases} 1/n & \text{if } \sigma = \sigma' \\ 2/n^2 & \text{if } \sigma' = \sigma \cdot (i, j) \text{ for some } 1 \leq i \neq j \leq n \\ 0 & \text{else} \end{cases}$$

Consider the following coupling argument for random transpositions. At each time  $t$ , select a label  $X_t$  uniformly at random in  $[n] = \{1, \dots, n\}$  and a position  $Y_t$  uniformly at random in  $[n]$ , independently. Then at time  $t$ , transpose the card with label  $X_t$  together with the card in position  $Y_t$ .

(a) Show that this mechanism generates the random transpositions Markov chain.

(b) By coupling show that  $t_{\text{mix}} = O(n^2)$  (consider the set of cards which occupy the same positions in the two decks).

(c) Let  $\sigma \in S_n$  be a random uniform permutation. Let  $X$  denote the number of fixed points of  $\sigma$ , i.e., the numbers of  $1 \leq i \leq n$  such that  $\sigma(i) = i$ . Show that  $\mathbb{E}(X) = 1$  and  $\text{Var}(X) = 1$ .

Deduce that for random transpositions,  $t_{\text{mix}} \geq (1/2 - \epsilon)n \log n$  for any  $\epsilon > 0$  and  $n$  sufficiently large.

6. The Random to Top shuffle (on  $n$  cards) is a random walk on  $S_n$  in which we have  $n$  distinct cards labeled  $[n] := \{1, \dots, n\}$ . At each step a random card is picked and is moved to the top of the deck. As a permutation, if  $X_t(i)$  is the label of the card at position  $i$  then  $X_{t+1} = X_t \circ (U_{t+1}, U_{t+1} - 1, \dots, 2, 1)$  where  $U_{t+1} \sim \text{Uniform}(\{1, \dots, n\})$ .

(a) Bound the mixing time using a coupling or using a strong stationary time.

(b) Bound the mixing time using the fact that  $P_{\text{RT}} = P_{\text{TR}}^T$ , where  $P_{\text{RT}}$  is the transition matrix of the Random to Top shuffle and  $P_{\text{TR}}$  is that of the Top to Random shuffle. Hint: Show that for each fixed  $t$  the two chains have the same  $d(t)$ .

7. (a) Let  $P$  be a transition matrix on state space  $S$  with a stationary distribution  $\pi$ . Recall that the time-reversal is given by  $P^*(x, y) = \frac{\pi(y)}{\pi(x)} P(y, x)$  for all  $x, y$ . Show that

$$\forall k \in \mathbb{N}, \quad \bar{d}(k) = \max_{f: S \rightarrow \mathbb{R}: \pi(f)=0, f \neq 0} \frac{\|(P^*)^k f\|_1}{\|f\|_1},$$

where  $\pi(f) := \sum_{x \in S} \pi(x)f(x)$  is the mean w.r.t.  $\pi$ ,  $\|f\|_1 := \pi(|f|)$  is the 1-norm w.r.t.  $\pi$  and  $P^*f(x) := \sum_y P^*(x, y)f(y)$ .

(Small hint:  $(P^*)^k = (P^k)^*$  and so one can reduce to the case  $k = 1$ .)

(HUGE HINT: try  $f = \frac{1_x}{\pi(x)} - \frac{1_y}{\pi(y)}$  and express general  $f$  of mean 0 as such a linear combination).

(b) Deduce that  $\bar{d}(t+s) \leq \bar{d}(t)\bar{d}(s)$ , for all  $s, t \in \mathbb{N}$ .

(c) Show that  $P$  is irreducible and aperiodic iff  $P^*$  is. (No need to submit solution to this part. Just check that you understand why this is the case.)

(d) Show that  $\lambda$  is an eigenvalue of  $P^*$  iff it is an eigenvalue of  $P$ .

(e) Show that if  $f$  is an eigenfunction of  $P$  with eigenvalue  $\lambda \neq 1$  then  $\pi(f) = 0$ .

(f) Deduce that if  $\lambda \neq 1$  is an eigenvalue of  $P$  then  $|\lambda|^t \leq \bar{d}(t)$  for all  $t$ . Hence  $|\lambda| \leq 1$ . Finally argue that if  $P$  is irreducible and aperiodic then  $|\lambda| < 1$ .

8. Let  $P$  is a transition matrix of an irreducible aperiodic Markov chain with stationary distribution  $\pi$ . The point of this exercise is to show that even when  $P$  is not reversible, some useful spectral-decomposition can be done in order to control the rate of decay of  $L_2$  distances from stationarity and of variances in terms of the absolute spectral-gap of  $PP^*$ . (However, in contrast with the reversible case, in general the distance from equilibrium cannot be bounded from below using the eigenvalues  $PP^*$ .)

(a) Show that  $Q := P^*P$  is reversible w.r.t.  $\pi$ .

(b) Assume that  $P(x, x) > 0$  for all  $x$ , and deduce that  $Q$  is aperiodic and irreducible.

(c) Denote the eigenvalues of  $Q$  by  $1 = \lambda_1(Q) > \lambda_2(Q) \geq \dots \geq \lambda_{|S|}(Q) > -1$ . Let  $\lambda_*(Q) := \max\{|\lambda_2(Q)|, |\lambda_{|S|}(Q)|\}$ . Show that for any function  $f : S \rightarrow \mathbb{R}$ ,

$$\text{Var}_\pi(Pf) \leq \lambda_*(Q) \text{Var}_\pi(f),$$

(where  $\text{Var}_\pi(g) := \pi[(g - \pi(g))^2]$ ) and deduce that for all  $k \in \mathbb{N}$

$$\text{Var}_\pi(P^k f) \leq \lambda_*(Q)^k \text{Var}_\pi(f).$$

(d) Show that for every initial distribution  $\mu$  if we denote  $\mu_k := \mu P^k$  (i.e.  $\mu_k(y) = \sum_z \mu(z)P^k(z, y)$  - this is the law of the discrete-time chain at time  $k$  if the initial distribution is  $\mu$ ) then  $\|\mu_k - \pi\|_{2,\pi}^2 = \text{Var}_\pi((P^*)^k f)$ , for  $f(x) := \frac{\mu(x)}{\pi(x)}$ .

(e) Deduce that  $\|\mu_k - \pi\|_{2,\pi}^2 \leq \lambda_*(PP^*)^k \|\mu - \pi\|_{2,\pi}^2$ .

9. Let  $P_t := e^{-t(I-P)}$ . The *Poincaré constant* of  $P$  is defined as  $\gamma := \min_{g:S \rightarrow \mathbb{R} \text{ non-constant}} \frac{\mathcal{E}(g,g)}{\text{Var}_\pi(g)}$ , where  $\mathcal{E}(g,g) := \langle (I-P)g, g \rangle_\pi$  (this is simply the spectral-gap when  $P$  is reversible). The point of this exercise is to show that even if  $P$  is not reversible,  $\gamma$  dictates the rate of decay of  $L_2$  distances and variances (in the continuous-time setup).

(a) Show that

$$\frac{d}{dt} \text{Var}_\pi(f_t) = -2\mathcal{E}(f_t, f_t),$$

where  $f_t := P_t f$  (hint:  $\frac{d}{dt} e^{tQ} = Q e^{tQ}$ ).

(b) Let  $\gamma := \min_{g: S \rightarrow \mathbb{R} \text{ non-constant}} \frac{\mathcal{E}(g, g)}{\text{Var}_\pi(g)}$ , where  $\mathcal{E}(g, g) := \langle (I - P)g, g \rangle_\pi$ . Deduce that

$$\forall t \geq 0, \quad \text{Var}_\pi(f_t) \leq e^{-2\gamma t} \text{Var}_\pi(f).$$

(c) Show that for every initial distribution  $\mu$  if we denote  $\mu_t := \mu P_t$  (i.e.  $\mu_t(y) = \sum_z \mu(z) P_t(z, y)$  - this is the law of the continuous-time chain at time  $t$  if the initial distribution is  $\mu$ ) then  $\|\mu_t - \pi\|_{2, \pi}^2 = \text{Var}_\pi(P_t^* f)$ , for  $f(x) := \frac{\mu(x)}{\pi(x)}$ , where  $P_t^* := e^{-t(I-P^*)}$ .

(make sure you understand why  $\pi(x)P_t^*(x, y) = \pi(y)P_t(y, x)$  for all  $x, y \in S$  and all  $t$ , and thus  $P_t = P_t^*$  for all  $t$  iff  $P = P^*$ .)

(d) Deduce that  $\|\mu_t - \pi\|_{2, \pi}^2 \leq e^{-2\gamma t} \|\mu - \pi\|_{2, \pi}^2$ .

10. Consider an irreducible, continuous-time reversible Markov chain. The goal of this exercise is to show that

$$t_{\text{sep}} := \inf\{t : \max_{x, y} 1 - \frac{P_t(x, y)}{\pi(y)} \leq 1/4\} \lesssim \min_x \max_y \mathbb{E}_y[T_x]. \quad (1)$$

The same holds in discrete-time if all of the eigenvalues of the transition matrix  $P$  are non-negative (e.g., if  $P$  is lazy, i.e.  $P(x, x) \geq 1/2$  for all  $x$ ).

(a)\*\* Let  $x, y, z$ . Let  $\tau_z^{(x)}$  and  $\tau_z^{(y)}$  be two independent random variables distributed like the hitting time of  $z$  started from  $x$  and  $y$  respectively. Show that for all  $t$  we have that

$$1 - P_t(x, y)/\pi(y) \leq \mathbb{P}[\tau_z^{(x)} + \tau_z^{(y)} > t].$$

You may use the fact that (in the continuous-time setup) the law of  $\tau_z^{(\cdot)}$  has a density function w.r.t. Lebesgue - make sure you understand why this is the case.) Please consider either the discrete time or continuous-time setups, but not both.

(b) Deduce (1).