

It is an intriguing, and nontrivial, exercise to work out the value of  $\lambda_2$  exactly in this case, in order to determine whether the mixing times  $\tau(\varepsilon)$  are closer to the given lower or upper bounds as a function of  $n$ .

Let us now return to the proof of Theorem 3.6, establishing the connection between the second-largest eigenvalue and the conductance of a Markov chain. Recall the statement of the Theorem:

**Theorem 3.6** *Let  $\mathcal{M}$  be a finite, regular, reversible Markov chain and  $\lambda_2$  the second-largest eigenvalue of its transition matrix. Then:*

- (i)  $\lambda_2 \leq 1 - \frac{\Phi^2}{2}$ ,
- (ii)  $\lambda_2 \geq 1 - 2\Phi$ .

*Proof.* (i) The approach here is to bound  $\Phi$  in terms of the eigenvalue gap of  $\mathcal{M}$ , i.e. to show that  $\Phi^2/2 \leq 1 - \lambda_2$ , from which the claimed result follows.

Thus, consider the eigenvalue  $\lambda = \lambda_2$ . (The following proof does not in fact depend on this particular choice of eigenvalue  $\lambda \neq 1$ , but since we are proving an upper bound of the form  $\Phi^2/2 \leq 1 - \lambda$ , all other eigenvalues yield weaker bounds than  $\lambda_2$ .)

Let  $e$  be a left eigenvector  $e \neq 0$  such that  $eP = \lambda e$ . Now  $e$  must contain both positive and negative components, since  $\sum_i e_i = 0$  as can be seen:

$$\begin{aligned} eP = \lambda e &\Leftrightarrow \sum_i e_i p_{ij} = \lambda e_j \quad \forall j \\ &\Rightarrow \sum_j \sum_i e_i p_{ij} = \sum_i e_i \underbrace{\sum_j p_{ij}}_{=1} = \lambda \sum_j e_j \\ &\stackrel{\lambda \neq 1}{\Rightarrow} \sum_i e_i = 0. \end{aligned}$$

Define  $A = \{i \mid e_i > 0\}$ . Assume, without loss of generality, that  $\pi(A) \leq 1/2$ . (Otherwise we may replace  $e$  by  $-e$  in the following proof.)

Define further a “ $\pi$ -normalised” version of  $e \upharpoonright A$ :

$$u_i = \begin{cases} e_i/\pi_i, & \text{if } i \in A \\ 0, & \text{if } i \notin A \end{cases}$$

Without loss of generality we may again assume that the states are indexed so that  $u_1 \geq u_2 \geq \dots \geq u_r > u_{r+1} = \dots = u_n = 0$ , where  $r = |A|$ .

In the remainder of the proof, the following quantity will be important:

$$D = \frac{\sum_{i<j} w_{ij}(u_i^2 - u_j^2)}{\sum_i \pi_i u_i^2}.$$

We shall prove the following claims:

- (a)  $\Phi \leq D$ ,
- (b)  $D^2/2 \leq 1 - \lambda$ ,

which suffice to establish our result.

*Proof of (a):* Denote  $A_k = \{1, \dots, k\}$ , for  $k = 1, \dots, r$ . The numerator in the definition of  $D$  may be expressed in terms of the ergodic flows out of the  $A_k$  as follows:

$$\begin{aligned} \sum_{i<j} w_{ij}(u_i^2 - u_j^2) &= \sum_{i<j} w_{ij} \sum_{i \leq k < j} (u_k^2 - u_{k+1}^2) \\ &= \sum_{k=1}^r (u_k^2 - u_{k+1}^2) \sum_{\substack{i \in A_k \\ j \notin A_k}} w_{ij} \\ &= \sum_{k=1}^r (u_k^2 - u_{k+1}^2) F_{A_k}. \end{aligned}$$

Now the capacities of the  $A_k$  satisfy  $\pi(A_k) \leq \pi(A) \leq 1/2$ , so by definition  $\Phi_{A_k} \geq \Phi \Rightarrow F_{A_k} \geq \Phi \cdot \pi(A_k)$ . Thus,

$$\begin{aligned} \sum_{i<j} w_{ij}(u_i^2 - u_j^2) &= \sum_{k=1}^r (u_k^2 - u_{k+1}^2) F_{A_k} \\ &\geq \Phi \cdot \sum_{k=1}^r (u_k^2 - u_{k+1}^2) \pi(A_k) \\ &= \Phi \cdot \sum_{k=1}^r (u_k^2 - u_{k+1}^2) \sum_{i=1}^k \pi_i \\ &= \Phi \cdot \sum_{i=1}^r \pi_i \sum_{k=i}^r (u_k^2 - u_{k+1}^2) \\ &= \Phi \cdot \sum_{i \in A} \pi_i u_i^2. \end{aligned}$$

Hence,

$$\Phi \leq \frac{\sum_{i<j} w_{ij}(u_i^2 - u_j^2)}{\sum_i \pi_i u_i^2} = D.$$

*Proof of (b):* We introduce one more auxiliary expression:

$$E = \frac{\sum_{i<j} w_{ij}(u_i - u_j)^2}{\sum_i \pi_i u_i^2},$$

and establish that: (b')  $D^2 \leq 2E$ , (b'')  $E \leq 1 - \lambda$ . This will conclude the proof of Theorem 3.6 (i).

*Proof of (b'):* Observe first that

$$\sum_{i<j} w_{ij}(u_i + u_j)^2 \leq 2 \sum_{i<j} w_{ij}(u_i^2 + u_j^2) \leq 2 \sum_{i \in A} \pi_i u_i^2.$$

Then, by the Cauchy-Schwartz inequality:

$$\begin{aligned} D^2 &= \left( \frac{\sum_{i<j} w_{ij}(u_i^2 - u_j^2)}{\sum_i \pi_i u_i^2} \right)^2 \\ &\leq \left( \frac{\sum_{i<j} w_{ij}(u_i + u_j)^2}{\sum_i \pi_i u_i^2} \right) \left( \frac{\sum_{i<j} w_{ij}(u_i - u_j)^2}{\sum_i \pi_i u_i^2} \right) \leq 2E. \end{aligned}$$

*Proof of (b''):* Denote  $Q = I - P$ . Then  $eQ = (1 - \lambda)e$  and thus

$$eQu^T = (1 - \lambda)eu^T = (1 - \lambda) \sum_{i=1}^r \pi_i u_i^2.$$

On the other hand, writing  $eQu^T$  out explicitly:

$$\begin{aligned}
 eQu^T &= \sum_{i=1}^n \sum_{j=1}^r q_{ij} e_i u_j & \left. \begin{aligned} q_{ij} &= -p_{ij} = -\frac{w_{ij}}{\pi_i}, \quad i \neq j \\ q_{ii} &= 1 - p_{ii} = \sum_{i \neq j} p_{ij} \\ e_i &= \pi_i u_i, \quad i \in A \end{aligned} \right| \\
 &\geq \sum_{i=1}^r \sum_{j=1}^r q_{ij} e_i u_j \\
 &= -\sum_{i \in A} \sum_{\substack{j \in A \\ j \neq i}} w_{ij} u_i u_j + \sum_{i \in A} \sum_{\substack{j \in A \\ j \neq i}} w_{ij} u_i^2 \\
 &= -2 \sum_{i < j} w_{ij} u_i u_j + \sum_{i < j} w_{ij} (u_i^2 + u_j^2) \\
 &= \sum_{i < j} w_{ij} (u_i - u_j)^2.
 \end{aligned}$$

Thus,

$$E \cdot \sum_i \pi_i u_i^2 = \sum_{i < j} w_{ij} (u_i - u_j)^2 \leq eQu^T = (1 - \lambda) \cdot \sum_i \pi_i u_i^2 \Rightarrow E \leq 1 - \lambda.$$

(ii) Given the stationary distribution vector  $\pi \in \mathbb{R}^n$ , define an inner product  $\langle \cdot, \cdot \rangle_\pi$  in  $\mathbb{R}^n$  as:

$$\langle u, v \rangle_\pi = \sum_{i=1}^n \pi_i u_i v_i.$$

By (a slight modification of) a standard result (the Courant-Fischer minimax theorem) in matrix theory, and the fact that  $P$  is reversible with respect to  $\pi$ , implying  $\langle u, Pv \rangle_\pi = \langle Pu, v \rangle_\pi$ , one can characterise the eigenvalues of  $P$  as:

$$\begin{aligned}
 \lambda_1 &= \max \left\{ \frac{\langle u, Pu \rangle_\pi}{\langle u, u \rangle_\pi} \mid u \neq 0 \right\}, \\
 \lambda_2 &= \max \left\{ \frac{\langle u, Pu \rangle_\pi}{\langle u, u \rangle_\pi} \mid u \perp \pi, u \neq 0 \right\}, \text{ etc.}
 \end{aligned}$$

In particular,

$$\lambda_2 \geq \frac{\langle u, Pu \rangle_\pi}{\langle u, u \rangle_\pi} \text{ for any } u \neq 0 \text{ such that } \sum_i \pi_i u_i = 0. \quad (5)$$

Given a set of states  $A \subseteq S$ ,  $0 < \pi(A) \leq 1/2$ , we shall apply the bound (5) to the vector  $u$  defined as:

$$u_i = \begin{cases} \frac{1}{\pi(A)}, & \text{if } i \in A \\ -\frac{1}{\pi(\bar{A})}, & \text{if } i \in \bar{A} \end{cases}$$

Clearly

$$\sum_i \pi_i u_i = \sum_{i \in A} \frac{\pi_i}{\pi(A)} - \sum_{i \in \bar{A}} \frac{\pi_i}{\pi(\bar{A})} = 1 - 1 = 0, \text{ and}$$

$$\langle u, u \rangle_\pi = \sum_i \pi_i u_i^2 = \sum_{i \in A} \frac{\pi_i}{\pi(A)^2} + \sum_{i \in \bar{A}} \frac{\pi_i}{\pi(\bar{A})^2} = \frac{1}{\pi(A)} + \frac{1}{\pi(\bar{A})},$$

so let us compute the value of  $\langle u, Pu \rangle_\pi$ .

The task can be simplified by representing  $P$  as  $P = I_n - (I_n - P)$ , and first computing  $\langle u, (I - P)u \rangle_\pi$ :

$$\begin{aligned} \langle u, (I - P)u \rangle_\pi &= \sum_i \pi_i u_i \sum_j (I - P)_{ij} u_j \\ &= - \sum_i \sum_{j \neq i} \pi_i u_i p_{ij} u_j + \sum_i \sum_{j \neq i} \pi_i u_i p_{ij} u_i \\ &= \sum_i \sum_{j \neq i} \pi_i p_{ij} (u_i^2 - u_i u_j) \\ &= \sum_{i < j} \pi_i p_{ij} (u_i - u_j)^2 \\ &= \sum_{\substack{i \in A \\ j \neq i}} \pi_i p_{ij} \left( \frac{1}{\pi(A)} + \frac{1}{\pi(\bar{A})} \right)^2 \\ &= \left( \frac{1}{\pi(A)} + \frac{1}{\pi(\bar{A})} \right)^2 F_A. \end{aligned}$$

Thus,

$$\begin{aligned} \lambda_2 &\geq \frac{\langle u, Pu \rangle_\pi}{\langle u, u \rangle_\pi} = \frac{1}{\langle u, u \rangle_\pi} \left( \langle u, u \rangle_\pi - \langle u, (I - P)u \rangle_\pi \right) \\ &= 1 - \frac{1}{\langle u, u \rangle_\pi} \cdot \langle u, (I - P)u \rangle_\pi \\ &= 1 - \left( \frac{1}{\pi(A)} + \frac{1}{\pi(\bar{A})} \right)^{-1} \left( \frac{1}{\pi(A)} + \frac{1}{\pi(\bar{A})} \right)^2 \cdot F_A \\ &= 1 - \left( \frac{1}{\pi(A)} + \frac{1}{\pi(\bar{A})} \right) \cdot F_A \\ &\geq 1 - \frac{2}{\pi(A)} \cdot F_A = 1 - 2\Phi_A. \end{aligned}$$

Since the bound (6) holds for any  $A \subseteq S$  such that  $0 < \pi(A) \leq 1/2$ , it follows that it holds also for the conductance

$$\Phi = \min_{0 < \pi(A) \leq 1/2} \Phi_A.$$

Thus, we have shown that  $\lambda_2 \geq 1 - 2\Phi$ , which completes the proof.  $\square$

Despite the elegance of the conductance approach, it can be sometimes (often?) difficult to apply in practice – computing graph conductance can be quite difficult. Also the bounds obtained are not necessarily the best possible; in particular the square in the upper bound  $\lambda_2 \leq 1 - \Phi^2/2$  is unfortunate.

An alternative approach, which is sometimes easier to apply, and can even yield better bounds, is based on the construction of so called “canonical paths” between states of a Markov chain.

Consider again a regular, reversible Markov chain with stationary distribution  $\pi$ , represented as a weighted graph with node set  $S$  and edge set  $E = \{(i, j) \mid p_{ij} > 0\}$ . The weight, or capacity,  $w_e$  associated to edge  $e = (i, j)$  corresponds to the ergodic flow  $\pi_i p_{ij}$  between states  $i$  and  $j$ .

Specify for each pair of states  $k, l \in S$  a *canonical path*  $\gamma_{kl}$  connecting them. The paths should intuitively be chosen as short and as nonoverlapping as possible. (For precise statements, see Theorems 3.9 and 3.11 below.)

Denote  $\Gamma = \{\gamma_{kl} \mid k, l \in S\}$  and define the unweighted and weighted *edge loading* induced by  $\Gamma$  on an edge  $e \in E$  as:

$$\begin{aligned} \rho_e &= \frac{1}{w_e} \sum_{\gamma_{kl} \ni e} \pi_k \pi_l \\ \bar{\rho}_e &= \frac{1}{w_e} \sum_{\gamma_{kl} \ni e} \pi_k \pi_l |\gamma_{kl}|, \end{aligned}$$

where  $|\gamma_{kl}|$  is the length (number of edges) of path  $\gamma_{kl}$ . (Note that here the edges are considered to be *oriented*, so that only paths crossing an edge  $e = (i, j)$  in the direction from  $i$  to  $j$  are counted in determining the loading of  $e$ .) The *maximum edge loading* induced by  $\Gamma$  is then:

$$\begin{aligned} \rho &= \rho(\Gamma) = \max_{e \in E} \rho_e \\ \bar{\rho} &= \bar{\rho}(\Gamma) = \max_{e \in E} \bar{\rho}_e. \end{aligned}$$

**Theorem 3.9** *For any regular, reversible Markov chain and any choice of canonical paths,*

$$\Phi \geq \frac{1}{2\rho}.$$

*Proof.* Represent the chain as a weighted graph  $G$ , where the weight (capacity) on edge  $e = (i, j)$  is defined as:

$$w_{ij} = \pi_i p_{ij} = \pi_j p_{ji}.$$

Every set of states  $A \subseteq S$  determines a cut  $(A, \bar{A})$  in  $G$ , and the conductance of the cut corresponds to its *relative capacity*:

$$\Phi_A = \frac{w(A, \bar{A})}{V_A} = \frac{1}{\pi(A)} \sum_{i \in A, j \in \bar{A}} w_{ij}.$$

Let then  $A$  be a set with  $0 < \pi(A) \leq \frac{1}{2}$  that minimises  $\Phi_A$ , and thus has  $\Phi_A = \Phi$ . Assume some choice of canonical paths  $\Gamma = \{\gamma_{ij}\}$ , and assign to each path  $\gamma_{ij}$  a “flow” of value  $\pi_i \pi_j$ . Then the total amount of flow crossing the cut  $(A, \bar{A})$  is

$$\sum_{i \in A, j \in \bar{A}} \pi_i \pi_j = \pi(A) \pi(\bar{A}),$$

but the cut edges, i.e. edges crossing the cut, have only total capacity  $w(A, \bar{A})$ . Thus, some cut edge  $e$  must have loading

$$\rho_e = \frac{1}{w_e} \sum_{\gamma_{ij} \ni e} \pi_i \pi_j \geq \frac{\pi(A) \pi(\bar{A})}{w(A, \bar{A})} \geq \frac{\pi(A)}{2w(A, \bar{A})} = \frac{1}{2\Phi}.$$

The result follows.  $\square$

**Corollary 3.10** *With notations and assumptions as above,*

$$\lambda_2 \leq 1 - \frac{1}{8\bar{\rho}^2}.$$

*Proof.* From Theorems 3.6 and 3.9.  $\square$

A more advanced proof yields a tighter result:

**Theorem 3.11** *With notations and assumptions as above:*

- (i)  $\lambda_2 \leq 1 - \frac{1}{\bar{\rho}}$
- (ii)  $\Delta(t) \leq \frac{(1 - 1/\bar{\rho})^t}{\min_{i \in A} \pi_i}$

$$(iii) \quad \tau(\varepsilon) \leq \bar{\rho} \left( \ln \frac{1}{\varepsilon} + \ln \frac{1}{\pi_{\min}} \right). \square$$

**Example 3.2** *Random walk on a ring.* Let us consider again the cyclic random walk of Figure 11. Clearly the stationary distribution is  $\pi = [\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n}]$ , and the ergodic flow on each edge  $e = (i, i \pm 1)$  is

$$w_e = \pi_i p_{i, i \pm 1} = \frac{1}{n} \cdot \frac{1}{4} = \frac{1}{4n}.$$

An obvious choice for a canonical path connecting nodes  $k, l$  is the shortest one, with length

$$|\gamma_{kl}| = \min\{|l - k|, n - |l - k|\}.$$

It is fairly easy to see that each (oriented) edge is now travelled by 1 canonical path of length 1, 2 of length 2, 3 of length 3,  $\dots$ ,  $\frac{n}{2}$  of length  $\frac{n}{2}$  (actually the last one is just an upper bound). Thus:

$$\begin{aligned} \bar{\rho} &= \max_e \frac{1}{w_e} \sum_{\gamma_{kl} \ni e} \pi_k \pi_l |\gamma_{ij}| \leq 4n \sum_{r=1}^{n/2} \frac{1}{n^2} \cdot r^2 \\ &= \frac{4}{n} \cdot \frac{1}{6} \cdot \frac{n}{2} \cdot \left( \frac{n}{2} + 1 \right) \cdot (n+1) = \frac{1}{6} (n+1)(n+2) \\ \Rightarrow \\ \tau(\varepsilon) &\leq \frac{1}{6} (n+1)(n+2) \left( \ln n + \ln \frac{1}{\varepsilon} \right) \\ &= \frac{1}{6} n^2 \left( \ln n + \frac{1}{\varepsilon} \right) + O\left(n \left( \ln n + \ln \frac{1}{\varepsilon} \right)\right). \end{aligned}$$

**Example 3.3** *Sampling permutations.* Let us consider the Markov chain whose states are all possible permutations of  $[n] = \{1, 2, \dots, n\}$ , and for any permutations  $s, t \in S_n$ :

$$p_{st} = \begin{cases} \frac{1}{2}, & \text{if } s = t, \\ \frac{1}{2} \cdot \binom{n}{2}^{-1}, & \text{if } s \text{ can be changed to } t \text{ by transposing two elements,} \\ 0, & \text{otherwise} \end{cases}$$

Thus, e.g. for  $n = 3$  we obtain the transition graph in Figure 12.

Clearly, the stationary distribution for this chain is  $\pi = [\frac{1}{n!}, \frac{1}{n!}, \dots, \frac{1}{n!}]$ , and the ergodic flow on each edge  $\tau = (s, t)$ , with  $s \neq t$ ,  $p_{st} > 0$ , is:

$$w_\tau = \pi_s p_{st} = \frac{1}{n!} \cdot \frac{1}{2} \cdot \binom{n}{2}^{-1}.$$

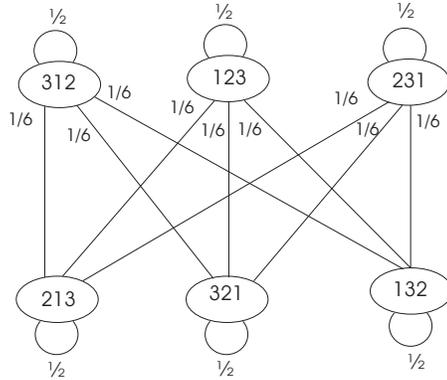


Figure 12: Transition graph for three-element permutations.

A natural canonical path connecting permutation  $s$  to permutation  $t$  is now obtained as follows:

$$s = s_0 \rightarrow s_1 \rightarrow s_2 \rightarrow \dots \rightarrow s_{n-1} = t.$$

where at each  $s_k, s_k(k) = t(k)$ . (Thus, each  $s_k$  matches  $t$  up to element  $k, s_k(1 \dots k) = t(1 \dots k)$ .)

Thus, e.g. the canonical path connecting  $s = (1234)$  to  $t = (3142)$  is as follows:

$$(1234) \rightarrow \overbrace{(3|214)}^{\omega} \xrightarrow{\tau} \overbrace{(31|24)}^{\omega'} \rightarrow (314|2).$$

Now let us denote the set of canonical paths containing a given transition  $\tau : \omega \rightarrow \omega'$  by  $\Gamma(\tau)$ . We shall upper bound the size of  $\Gamma(t)$  by constructing an injective mapping  $\sigma_\tau : \Gamma(\tau) \rightarrow S_n$ . Obviously, the existence of such a mapping implies that  $|\Gamma(\tau)| \leq n!$ .

Suppose  $\tau$  transposes locations  $k+1$  and  $l, k+1 < l$ , of permutation  $\omega$ . Then for any  $\langle s, t \rangle \in \Gamma(\tau)$ , define the permutation  $z = \sigma_\tau(s, t)$  as follows:

1. Place the elements in  $\omega(1 \dots k)$  in the locations they appear in  $s$ . (Note that permutation  $\omega$  is given and fixed as part of  $\tau$ .)
2. Place the remaining elements in the remaining locations in the order they appear in  $t$ .

Thus, for example in the above example case:

$$\sigma_\tau(\langle (1234), (3142) \rangle) \rightarrow (- \quad - \quad 3 \quad -) \rightarrow \underbrace{(1432)}_z$$

$$\omega = (3|214), \quad k = 1$$

Why is this mapping an injection, i.e. how do we recover  $s$  and  $t$  from a knowledge of  $\tau$  and  $z = \sigma_\tau(s, t)$ ? The reasoning goes as follows:

1.  $t = \omega(1 \dots k) +$  “other elements in same order as in  $z$ ”
2.  $s =$  “elements in  $\omega(1 \dots k)$  at locations indicated in  $z$ ” + “other elements in locations deducible from the transposition path  $s = s_0 \rightarrow s_1 \rightarrow \dots \rightarrow s_k = \omega$ ”

This is somewhat tricky, so let us consider an example. Say  $\omega = (3 \ 1|2 \ 4)$ ,  $k = 2$ ,  $z = (1 \ 4 \ 3 \ 2)$ . Then:

1.  $t = (3 \ 1|_ - \ -) + (- \ -|4 \ 2) = (3 \ 1|4 \ 2)$
- 2.

$$\begin{array}{rcl}
 s & = & s_0 = (1 \ - \ 3 \ -) \\
 & & s_1 = (3| \ - \ - \ -) \\
 \omega & = & s_2 = (3 \ 1| \ 2 \ 4) \\
 \hline
 \therefore s & = & s_0 = (1 \ 2 \ 3 \ 4) \\
 & & s_1 = (3| \ 2 \ 1 \ 4) \\
 \omega & = & s_2 = (3 \ 1| \ 2 \ 4)
 \end{array}
 \quad \Rightarrow \quad
 \begin{array}{rcl}
 s_0 & = & (1 \ - \ 3 \ -) \\
 s_1 & = & (3| \ 2 \ 1 \ -) \\
 s_2 & = & (3 \ 1| \ 2 \ 4) \\
 s_0 & = & (1 \ 2 \ 3 \ 4) \\
 s_1 & = & (3| \ 2 \ 1 \ 4) \\
 s_2 & = & (3 \ 1| \ 2 \ 4)
 \end{array}$$

Thus, we know that for each transition  $\tau$ ,

$$|\Gamma(\tau)| \leq n!$$

We can now obtain a bound on the unweighted maximum edge loading induced by our collection of canonical paths:

$$\begin{aligned}
 \rho &= \max_{\tau \in E} \frac{1}{q_\tau} \sum_{\langle s, t \rangle \in \Gamma(\tau)} \pi_s \pi_t \leq \left( \frac{1}{n!} \cdot \frac{1}{2} \cdot \binom{n}{2}^{-1} \right)^{-1} \cdot n! \cdot \left( \frac{1}{n!} \right)^2 \\
 &= 2n! \binom{n}{2} \cdot n! \cdot \left( \frac{1}{n!} \right)^2 = 2 \cdot \binom{n}{2} = n(n-1).
 \end{aligned}$$

By Theorem 3.9, the conductance of this chain is thus  $\Phi \geq \frac{1}{2n(n-1)}$ , and by Corollary 3.8, its mixing time is thus bounded by

$$\begin{aligned}
 \tau_n(\varepsilon) &\leq \frac{2}{\Phi^2} \left( \ln \frac{1}{\varepsilon} + \ln \frac{1}{\pi_{\min}} \right) \leq 2(2n(n-1))^2 \left( \ln \frac{1}{\varepsilon} + \ln n! \right) \\
 &= O \left( n^4 \left( n \ln n + \ln \frac{1}{\varepsilon} \right) \right).
 \end{aligned}$$