

# MATH4240 Tutorial 8 Notes

## 1 Periodicity

- The *period*  $d_x$  of a state  $x$  in a Markov chain is the gcd of all  $t$  that  $x$  can return to  $x$  after  $t$  steps,

$$d_x = \gcd \{ t \geq 1 \mid P^t(x, x) > 0 \}$$

and  $\{ t \geq 1 \mid P^t(x, x) > 0 \}$  contains all but finitely many multiplies of  $d_x$ .

(If the set is empty, the period is undefined or infinity, depending on your convention.)

- If  $x \rightarrow y$  and  $y \rightarrow x$ , then  $d_x = d_y$ .

In particular, in an irreducible closed set, every state has the same period

- An irreducible chain is called *aperiodic* / *acyclic* if the period is 1

Recall that a transition matrix is called *primitive* if there exists  $N \geq 1$  such that every entry in  $P^N$  is positive for all  $n \geq N$ . The following lemma is easy to show:

**Lemma 1.1.** *The transition matrix of a finite irreducible chain is primitive if and only if the chain is aperiodic.*<sup>1</sup>

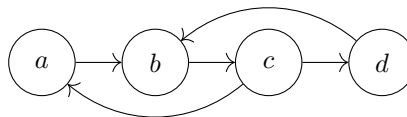
As noted before, a primitive matrix satisfies the conditions of the theorem from the lecture, so we have the following result:

**Theorem 1.2.** *If  $P$  is the transition matrix of a finite **irreducible aperiodic** chain, then  $\lim_n P^n = (\pi \ \dots \ \pi)^T$  where  $\pi$  is the unique normalized left eigenvector of  $P$  associated to the eigenvalue 1, which is also the unique stationary distribution of the chain.*

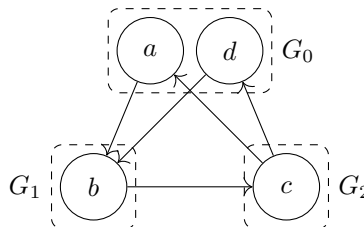
Together with the result on positive recurrence,

**Corollary 1.3.** *A finite **irreducible aperiodic** Markov chain is **positive recurrent**.*

*Example 1.* Consider the chain with transition diagram (ignoring the transition probabilities)



As noted in lecture, this chain has period  $d = 3$ . It may be easier to notice if we draw the diagram in this way:



Notice that by partitioning the state space into  $G_0, \dots, G_{d-1} = G_2$ , every state in  $G_i$  can only go to some states in  $G_{i+1}$  (and  $G_{d-1}$  to  $G_0$ ), and there is no transition to states in other classes.

<sup>1</sup>An irreducible aperiodic Markov chain is sometimes also called *ergodic*.

In fact, it is easy to show the following result (you should try to prove it):

**Theorem 1.4.** Consider the binary relation  $\sim$  on the state space of an irreducible chain with transition matrix  $P$  where  $x \sim y$  if  $P^{nd}(x, y) \neq 0$  for some  $n \geq 1$ . Then it is an equivalence relation that satisfies the following properties:

- there are exactly  $d$  equivalence classes (**cyclic classes**)
- the cyclic classes can be labeled as  $G_0, \dots, G_{d-1}$  in a way that, in one step, a state in  $G_i$  can only reach another state in  $G_{i+1}$  (identifying  $G_d = G_0$ )
- the transition matrix can be rearranged into the following **cyclic block form**<sup>2</sup>

$$P_{\text{cyc}} = \begin{matrix} & \begin{matrix} G_0 & G_1 & G_2 & \dots & G_{d-1} \end{matrix} \\ \begin{matrix} G_0 \\ G_1 \\ G_2 \\ \vdots \\ G_{d-1} \end{matrix} & \begin{bmatrix} 0 & P_1 & 0 & \dots & 0 \\ 0 & 0 & P_2 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ P_0 & 0 & 0 & \dots & 0 \end{bmatrix} \end{matrix}$$

for some matrices  $P_0, \dots, P_{d-1}$ .

- $P_{\text{cyc}}^d$  is a block diagonal matrix and each diagonal block is primitive

By definition, each of  $P_i$  is a (row) stochastic matrix and so satisfies  $P_i \vec{1} = \vec{1}$  (with appropriate sizes). This implies, on

$$v = (\vec{1} \quad \omega \vec{1} \quad \omega^2 \vec{1} \quad \dots \quad \omega^{d-1} \vec{1})^\top$$

with  $\omega \in \mathbb{C}$  being a  $d$ th root of unity ( $\omega^d = 1$ ), and  $\vec{1}$  vectors having appropriate sizes,

$$\begin{aligned} Pv &= (P_1 \omega \vec{1} \quad P_2 \omega^2 \vec{1} \quad \dots \quad P_{d-1} \omega^{d-1} \vec{1} \quad P_0 \vec{1})^\top \\ &= (\omega \vec{1} \quad \omega^2 \vec{1} \quad \dots \quad \omega^d \vec{1})^\top \\ &= \omega v \end{aligned}$$

so  $\omega$  is an eigenvalue of  $P$ .

As the diagonal blocks of  $P_{\text{cyc}}^d$  are primitive, from the result mentioned in lecture (also Tutorial 6), we can see that, if  $\lambda \in \mathbb{C}$  is an eigenvalue of  $P$  with modulus 1,  $\lambda^d$  must be an eigenvalue with modulus 1 to each of the diagonal block of  $P_{\text{cyc}}^d$ , and so  $\lambda^d = 1$ .

This gives us the following result<sup>3</sup>:

**Theorem 1.5** (Nonnegative Perron–Frobenius theorem, partial). If  $P$  is the transition matrix of a finite irreducible chain, then  $P$  has period  $d \geq 1$  if and only if  $P$  has exactly  $d$  eigenvalues with unit modulus, and in such case they are exactly the  $d$ th roots of unity:  $e^{2\pi i \frac{1}{d}}, \dots, e^{2\pi i \frac{d}{d}} = 1$ .

So now you have another way to compute the period of an irreducible chain<sup>4</sup>: instead of finding all loops or drawing the diagram in a clever way, just find *all* eigenvalues of the transition matrix and count those with unit modulus. However, note that sometimes it *may* be easier to just count the paths than to find all eigenvalues (especially when you have many states).

<sup>2</sup>Like canonical form, it seems that there isn't a common notation for this.

<sup>3</sup>For more, see this note by M. Boyle.

<sup>4</sup>There is also an algorithmic one, just like how irreducible closed sets can be found algorithmically. See this note by Jarvis and Shier for a BFS approach (following Denardo's result) and a corresponding Python implementation.

## 2 With Mean Recurrence Time

Recall from the last tutorial that the following holds on an irreducible chain:

$$\begin{aligned}\frac{1}{n}N_n(y) &\xrightarrow{n \rightarrow \infty} 1/m_y \quad \text{with probability 1} \\ \frac{1}{n}E_x(N_n(y)) &\xrightarrow{n \rightarrow \infty} 1/m_y\end{aligned}$$

Suppose the chain has period  $d$ , and  $x \in S$  be a state. As the label is not important, we may assume that  $x \in G_0$ . Then for each  $y \in G_r$ ,  $P^n(x, y) \neq 0$  only if  $n \equiv r \pmod{d}$ . If we group the terms together, for  $n = kd$ ,

$$\frac{1}{n}N_n(y) = \frac{1}{kd} \sum_{t=1}^{kd} \mathbf{1}_y(X_t) = \frac{1}{kd} \sum_{m=0}^{k-1} \mathbf{1}_y(X_{md+r}) \xrightarrow{n \rightarrow \infty} 1/m_y \quad \text{with probability 1}$$

and so

$$\begin{aligned}\frac{1}{k} \sum_{m=0}^{k-1} \mathbf{1}_y(X_{md+r}) &\xrightarrow{k \rightarrow \infty} d/m_y \quad \text{with probability 1} \\ \frac{1}{k} \sum_{m=0}^{k-1} P^{md+r}(x, y) &\xrightarrow{k \rightarrow \infty} d/m_y\end{aligned}$$

In particular,

$$\frac{1}{k} \sum_{m=0}^{k-1} P^{md}(x, x) \xrightarrow{k \rightarrow \infty} d/m_x$$

Of course, these are obvious once you have the strong convergence  $\lim_m P^{md+r}$  (see lecture notes).

Suppose for the moment the chain is positive recurrent, and so has unique stationary distribution  $\pi(x) = 1/m_x$ , and the transition matrix is already in cyclic block form. Let  $Q_0, \dots, Q_{d-1}$  be the diagonal blocks of  $P^d$ , so  $P^d = \text{diag}(Q_0, \dots, Q_{d-1})$ . As noted, each  $Q_i$  is primitive and so the corresponding  $d$ -step chain has its own (unique) stationary distribution  $\pi_i$  on  $G_i$ . In particular, as  $x \in G_0$ ,

$$\frac{1}{k} \sum_{m=0}^{k-1} P^{md}(x, x) = \frac{1}{k} \sum_{m=0}^{k-1} Q_0^m(x, x) \xrightarrow{k \rightarrow \infty} 1/m_{x, Q_0} = \pi_0(x)$$

Comparing with the previous limit, we have

$$\pi_0(x) = d/m_x = d\pi(x)$$

that is,

$$\pi(x) = \pi_0(x)/d$$

As  $x$  is arbitrary,

**Theorem 2.1.** *The stationary distribution of the original chain with transition matrix  $P$  is formed by putting together the stationary distributions on the cyclic classes of the  $d$ -step chain (with transition matrix  $P^d$ ) and normalizing by the period:*

$$\pi = (\pi_0, \pi_1, \dots, \pi_{d-1})/d$$

and, with  $P_i$  from the cyclic block form,

$$\pi_i P_{i+1} = \pi_{i+1}$$

Of course, this can be proven by inspecting the transition matrices (instead of going through the limits). This also means that, to find the stationary distribution of a periodic chain, you just need to find the stationary distribution of the  $d$ -step chain on *one* cyclic class, which could be easier to solve than the whole stationary equation (with the additional cost of finding cyclic classes,  $d$ -step matrix, and computing the matrix multiplications).