

**Lecture 7. More about discrete Markov chains.**

Considering periodicity is simpler than the rest, because it doesn't have anything to do with the limits or infinite series. But first of all we solve the following problem:

**7**  $f_{yy}^{(k)} = 0$  for all  $k$  that are not multiples of  $d$  if and only if  $p_{yy}^{(n)} = 0$  for all  $n$  that are not multiples of  $d$ .

We defined periodicity of a state through the first-return probabilities  $f_{yy}^{(k)}$ ; this problem shows that we can define it through the transition probabilities  $p_{yy}^{(n)}$ .

**Theorem 7.1.** *If one state  $y$  belonging to an equivalence class  $C_i$  is periodic with period  $d$ , then all other states of this class are periodic with period  $d$ ; if one state is aperiodic, then all other states of the same class are aperiodic too.*

**Proof.** Suppose we can return from  $y$  to  $y$  only in a number of steps that is a multiple of  $d$ ; and suppose that we can return from  $x$  to  $x$  in a number of steps  $N$  that is *not* a multiple of  $d$ . We can travel from  $y$  to  $x$  in  $K$  steps:  $p_{yx}^{(K)} > 0$ , and return to  $y$  from  $x$  in  $M$  steps:  $p_{xy}^{(M)} > 0$ . Then we can return from  $y$  to  $y$  in  $K + M$  steps:  $p_{yy}^{(K+M)} \geq p_{yx}^{(K)} \cdot p_{xy}^{(M)} > 0$ . As we assumed,  $K + M$  is a multiple of  $d$ . But we can also return from  $y$  to the same point in  $K + N + M$  steps: first going from  $y$  to  $x$  in  $K$  steps, then from  $x$  to  $x$  in  $N$ , and finally from  $x$  to  $y$  in  $M$  steps. But since  $N$  is not a multiple of  $d$ , the sum  $K + N + M$  also isn't one. We get a contradiction, whose source is that we supposed that  $x$  and  $y$  had different periods. So this is impossible.

**Theorem 7.2.** *The states belonging to the same equivalence class  $C_i$  are either all transient, or all recurrent. If  $\lim_{n \rightarrow \infty} p_{yy}^{(n)} = 0$  for some  $y \in C_i$ , then  $\lim_{n \rightarrow \infty} p_{xx}^{(n)} = 0$  for all states  $x$  from the same equivalence class  $C_i$ . If the limit  $\lim_{n \rightarrow \infty} p_{yy}^{(n)} = p_y > 0$ , then  $\lim_{n \rightarrow \infty} p_{xy}^{(n)} = p_y$  for all  $x$  of the same equivalence class, and  $\lim_{n \rightarrow \infty} p_{xx}^{(n)} = p_x$  exists and is positive.*

**Proof.** Let  $x \sim y$ . If we have  $p_{yx}^{(N)} > 0, p_{xy}^M > 0$ , then for  $n \geq N + M$  we have:

$$p_{yy}^{(n)} = P_y\{\xi_n = y\} \geq P_y\{\xi_N = x, \xi_n = y\} = p_{yx}^{(N)} \cdot p_{xy}^{(n-N)} = \text{const} \cdot p_{xy}^{(n-N)}, \tag{7.1}$$

$$p_{yy}^{(n)} \geq P_y\{\xi_{n-M} = x, \xi_n = y\} = p_{yx}^{(n-M)} \cdot p_{xy}^{(M)} = \text{const} \cdot p_{yx}^{(n-M)}, \tag{7.2}$$

$$p_{yy}^{(n)} \geq p_{yx}^{(N)} \cdot p_{xx}^{(n-N-M)} \cdot p_{xy}^{(M)} = \text{const} \cdot p_{xx}^{(n-N-M)} \tag{7.3}$$

with positive constants. So if the series  $\sum_{n=0}^{\infty} p_{yy}^{(n)}$  converges, so do the series  $\sum_{n=N}^{\infty} p_{xy}^{(n)}, \sum_{n=M}^{\infty} p_{yx}^{(n)}, \sum_{n=N+M}^{\infty} p_{xx}^{(n)}$ , which is equivalent to convergence of  $\sum_{n=0}^{\infty} p_{xy}^{(n)}, \sum_{n=0}^{\infty} p_{yx}^{(n)}$ , and  $\sum_{n=0}^{\infty} p_{xx}^{(n)}$ .

Also if  $\lim_{n \rightarrow \infty} p_{yy}^{(n)} = 0$ , also are equal to 0 the limits  $\lim_{n \rightarrow \infty} p_{xy}^{(n)}, \lim_{n \rightarrow \infty} p_{yx}^{(n)}$ , and  $\lim_{n \rightarrow \infty} p_{xx}^{(n)} = 0$ ; and if  $\lim_{n \rightarrow \infty} p_{yy}^{(n)} > 0$ , the limit  $\lim_{n \rightarrow \infty} p_{xx}^{(n)}$  cannot be equal to 0.

Now suppose  $\lim_{n \rightarrow \infty} p_{yy}^{(n)} = p_y > 0$ . Then the state  $y$  (and with it all other states  $x$  in the equivalence class  $C_i$ ) cannot be transient, and it cannot be periodic. By Theorem 6.2 we have  $P_x\{\tau_y < \infty\} = 1$ , and by Theorem 6.1 we have  $\lim_{n \rightarrow \infty} p_{xy}^{(n)} = p_y$ .

Is our Markov chain, if we consider it on a recurrent aperiodic class  $C_i$ , necessarily ergodic? That is: do the limiting probabilities  $p_y = \lim_{n \rightarrow \infty} p_{xy}^{(n)}$  form a probability distribution: is  $\sum_{y \in C_i} p_y = 1$ ?

In the case of a *finite*  $C_i$ , yes:  $\sum_{y \in C_i} p_y = \sum_{y \in C_i} \lim_{n \rightarrow \infty} p_{xy}^{(n)} = \lim_{n \rightarrow \infty} \sum_{y \in C_i} p_{xy}^{(n)} = \lim_{n \rightarrow \infty} 1 = 1$ . In the case of (countably) infinite class, it is possible that  $p_y = \lim_{n \rightarrow \infty} p_{xy}^{(n)} = 0$  for all  $y \in C_i$  (as in the case of the simplest random walk – which is, though, a periodic chain with period 2). But if all  $p_y > 0$ , is their sum necessarily equal to 1? can it be equal, say, to  $2/3$  or to  $3/2$ ?

**Theorem 7.3.** *For the limits  $p_y = \lim_{n \rightarrow \infty} p_{xy}^{(n)}$  (supposing these limits exist), we have:  $\sum_{y \in C_i} p_y \leq 1$ .*

**Proof.** Let us number the points of the class  $C_i$ :  $x_1, x_2, \dots, x_j, \dots$  (we don't have to consider the case of  $C_i$  being finite). Let  $x_0$  be a fixed state belonging to the class  $C_i$  ( $x_0$  is equal to one of  $x_1, x_2, \dots, x_j, \dots$ ). We have:

$$\sum_{j=1}^m p_{x_j} = \sum_{j=1}^m \lim_{n \rightarrow \infty} p_{x_0 x_j}^{(n)} = \lim_{n \rightarrow \infty} \sum_{j=1}^m p_{x_0 x_j}^{(n)} \leq \lim_{n \rightarrow \infty} \sum_{y \in C_i} p_{x_0 y}^{(n)} = 1, \quad (7.4)$$

$$\sum_{y \in C_i} p_y = \lim_{m \rightarrow \infty} \sum_{j=1}^m p_{x_j} \leq 1. \quad (7.5)$$

**Theorem 7.4.** *For all  $y \in C_i$*

$$\sum_{x \in C_i} p_x \cdot p_{xy} = p_y \quad (7.6)$$

(in matrix notations:  $\mathbf{p} \cdot P = \mathbf{p}$ , where  $\mathbf{p}$  is the row vector with components  $p_y, y \in C_i$ , and  $P = (p_{xy})_{x, y \in C_i}$  is the one-step transition matrix restricted to the class  $C_i$ ).

**Proof.** Take a finite sum first:

$$\begin{aligned} \sum_{j=1}^m p_{x_j} \cdot p_{x_j y} &= \sum_{j=1}^m \lim_{n \rightarrow \infty} p_{x_0 x_j}^{(n)} \cdot p_{x_j y} = \lim_{n \rightarrow \infty} \sum_{j=1}^m p_{x_0 x_j}^{(n)} \cdot p_{x_j y} \\ &\leq \lim_{n \rightarrow \infty} \sum_{x \in C_i} p_{x_0 x}^{(n)} \cdot p_{xy} = \lim_{n \rightarrow \infty} p_{x_0 y}^{(n+1)} = p_y. \end{aligned} \quad (7.7)$$

Taking  $m \rightarrow \infty$ , we get:

$$\sum_{x \in C_i} p_x \cdot p_{xy} = \lim_{m \rightarrow \infty} \sum_{j=1}^m p_{x_j} \cdot p_{x_j y} \leq p_y. \quad (7.8)$$

Now, suppose that a strict inequality holds in (7.8) for some  $y = y_0$ :

$$\sum_{x \in C_i} p_x \cdot p_{xy_0} < p_{y_0}. \quad (7.9)$$

Let us take the sum of both sides in (7.8) over all  $y \in C_i$ . Since all left-hand sides are not greater than the corresponding right-hand sides, and at least one is smaller, we have:

$$\sum_{y \in C_i} \sum_{x \in C_i} p_x \cdot p_{xy} < \sum_{y \in C_i} p_y. \quad (7.10)$$

Of course, when we see an iterated (a double) sum (and the summands are nonnegative), we should reverse the order of summation:

$$\sum_{x \in C_i} \sum_{y \in C_i} p_x \cdot p_{xy} < \sum_{y \in C_i} p_y. \quad (7.11)$$

The sum in the left-hand side is equal to

$$\sum_{x \in C_i} p_x \sum_{y \in C_i} p_{xy} = \sum_{x \in C_i} p_x \cdot 1 = \sum_{x \in C_i} p_x. \quad (7.12)$$

So we get  $\sum_{x \in C_i} p_x < \sum_{y \in C_i} p_y$ : a contradiction. Its source is our assuming (7.9); so it's impossible, and equality (7.6) holds for all  $y$ .

**Theorem 7.5.** *If  $p_y = \lim_{n \rightarrow \infty} p_{xy}^{(n)} > 0$ , then  $\sum_{y \in C_i} p_y = 1$ : ergodicity.*

**Proof.** By Theorem 7.4, we have  $\mathbf{p} \cdot P = \mathbf{p}$ . Multiplying both parts by  $P$  on the right, we get:  $\mathbf{p} \cdot P^2 = \mathbf{p} \cdot P = \mathbf{p}$ . Multiplying again and again by  $P$ , we get  $\mathbf{p} \cdot P^n = \mathbf{p}$  for every  $n$ . In components, it is written as

$$\sum_{x \in C_i} p_x \cdot p_{xy}^{(n)} = p_y. \quad (7.13)$$

Now we take  $n \rightarrow \infty$ . The  $x$ -th term of the sum converges to  $p_x \cdot p_y$  as  $n \rightarrow \infty$ ; but does the sum converge to  $\sum_{x \in C_i} p_x \cdot p_y$ ? (This is an *infinite* sum, and the limit passage needs justification.) We are using the dominated-convergence theorem (remember, a sum is a Lebesgue integral with respect to the counting measure  $\#$ ): the summands  $p_x \cdot p_{xy}^{(n)}$  are, for every  $n$ , dominated by the function  $p_x$ , and  $\sum_{x \in C_i} p_x \leq 1$  by Theorem 7.3. So we have

$$p_y = \lim_{n \rightarrow \infty} \sum_{x \in C_i} p_x \cdot p_{xy}^{(n)} = \sum_{x \in C_i} \lim_{n \rightarrow \infty} p_x \cdot p_{xy}^{(n)} = \sum_{x \in C_i} p_x \cdot p_y. \quad (7.14)$$

Since  $p_y > 0$ , we can divide both sides by  $p_y$  and get:  $\sum_{x \in C_i} p_x = 1$ .

For an aperiodic class  $C_i$  with  $p_y > 0$ ,  $y \in C_i$ , we have:

$$\sum_{y \in C_i} \frac{1}{E_y \tau_y} = 1; \quad (7.15)$$

and  $\lim_{n \rightarrow \infty} p_{xy}^{(n)} = p_y$  for every  $x \sim y$ .

What can we say about  $\lim_{n \rightarrow \infty} p_{xy}^{(n)}$  for  $x \not\sim y$ ? This means that either we can go from  $y$  to  $x$ , but not back; or that  $x \rightarrow y$ , but we cannot go from  $y$  to  $x$ ; or that we can neither do from  $y$  to  $x$ , nor from  $x$  to  $y$ .

In the first and in the last case clearly  $p_{xy}^{(n)} = 0$ ,  $\lim_{n \rightarrow \infty} p_{xy}^{(n)} = 0$ ; but the second case is possible – if the state  $x$  is transient. In this case, as we know (Theorem 6.1),  $\lim_{n \rightarrow \infty} p_{xy}^{(n)} = P_x\{\tau_y < \infty\} \cdot p_y$ ; but how to find  $P_x\{\tau_y < \infty\}$ ??

But it seems we have come closer to the end of our considering discrete Markov chains; let me drop this question for the time being.

I want to say a couple of words about the *periodic* case.

In this case we can consider on the equivalence class  $C_i$  the aperiodic Markov chain  $\tilde{\xi}_n = \xi_{nd}$  with one-step transition matrix  $\tilde{P} = P^d$  and  $\tilde{p}_{xy}^{(n)} = p_{xy}^{(nd)}$ . We saw (Theorem 5.3) that  $\lim_{n \rightarrow \infty} p_{yy}^{(nd)} = 1/E_y \tilde{\tau}_y = d/E_y \tau_y$ , and  $p_{yy}^{(k)} = 0$  for  $k$  not being multiples of  $d$ .

We can subdivide the equivalence class  $C_i$  into  $d$  subclasses in the following way. We introduce a new equivalence relation  $\sim_d$ :  $x \sim_d y$  if we can go from  $x$  to  $y$  and from  $y$  to  $x$  in some number of steps that is a multiple of  $d$ :  $x \sim_d y$  if there exists an integer  $n$  such that  $p_{xy}^{(nd)} > 0$ , and another integer  $m$  such that  $p_{yx}^{(md)} > 0$ . For some  $x_0 \in C_i$  we introduce the class of  $\sim_d$ -equivalence  $C_{i0} \subset C_i$  containing  $x_0$ :  $C_{i0} = \{x: x \sim_d x_0\}$ .

Then we consider the subclass  $C_{i1}$  consisting of all states  $z \in C_i$  such that the one-step transition probability  $p_{xz} > 0$  for some  $x \in C_{i0}$ . We can reach a point  $z \in C_{i1}$  from the point  $x_0$  (and from every other point  $x \in C_{i0}$ ) only in number of steps  $n$  that has the form  $n = 1$ , or  $n = d + 1$ , or  $n = 2d + 1$ , ...: all  $n$  with remainder 1 after division by  $d$ . Then we introduce  $C_{i2} = \{z \in C_i: p_{x_0 z}^{(n)} = 0 \text{ for } n = 2, d + 2, 2d + 2, 3d + 2, \dots\}$ , etc., and finally  $C_{i, d-1} = \{z \in C_i: p_{x_0 z}^{(n)} = 0 \text{ except for } n = d - 1, 2d - 1, 3d - 1, \dots\}$ .

It is easy to check that  $C_i = \bigcup_{j=0}^{d-1} C_{ij}$ , that the subclasses  $C_{ij}$  are disjoint for different  $j$ , and that every  $x$  and  $y$  belonging to the same subclass are  $\sim_d$ -equivalent. If we start from a state  $\xi_0$  belonging to a subclass  $C_{ij}$ , our next state  $\xi_1$  is almost surely in the subclass  $C_{i, j+1}$  (if  $j \neq d - 1$ ; if  $j = d - 1$ , it is in the subclass  $C_{i0}$ ); the next state  $\xi_2$  is in  $C_{i, j+2}$  or  $C_{i, j+2-d}$ ; etc.

For  $x \in C_{ij}$ ,  $y \in C_{ik}$  we have  $\lim_{n \rightarrow \infty} p_{xy}^{(nd+m)} = d/E_y \tau_d$  if  $m + j - k$  is divisible by  $d$ , and  $p_{xy}^{(nd+m)} = 0$  if it is not.

The sum  $\sum_{y \in C_{ij}} d/E_y \tau_y = 1$ , so  $\sum_{y \in C_{ij}} 1/E_y \tau_y = 1/d$ , and  $\sum_{y \in C_i} 1/E_y \tau_y = 1$ .