11.2.4 Classification of States

To better understand Markov chains, we need to introduce some definitions. The first definition concerns the accessibility of states from each other: If it is possible to go from state ii to state jj, we say that

state jj is *accessible* from state ii. In particular, we can provide the following definitions.

We say that state jj is **accessible** from state ii, written as $i \rightarrow ji \rightarrow j$, if $p_{(n)ij} > 0$ pij(n) > 0 for some nn. We assume every state is accessible from itself since $p_{(0)ii} = 1$ pii(0) = 1.

Two states ii and jj are said to **communicate**, written as $i \leftrightarrow j i \leftrightarrow j$, if they are **accessible** from each other. In other words,

$$i \leftrightarrow j \text{ means } i \rightarrow j \text{ and } j \rightarrow i. i \leftrightarrow j \text{ means } i \rightarrow j \text{ and } j \rightarrow i.$$

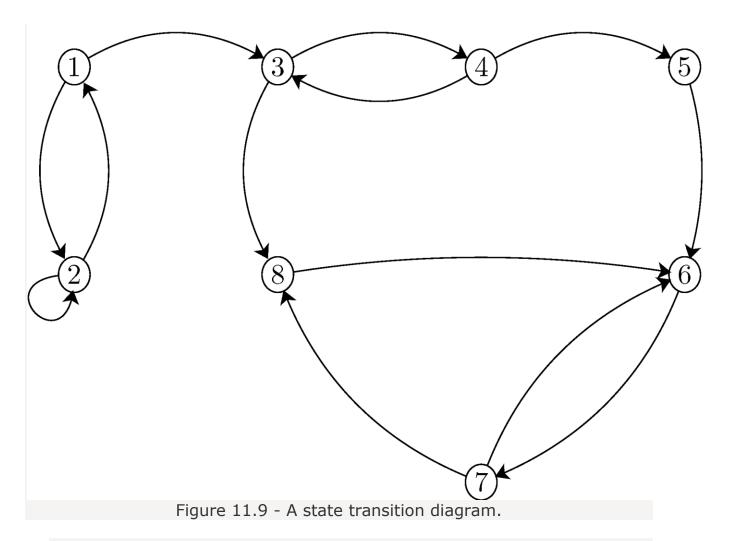
Communication is an equivalence relation. That means that

- -every state communicates with itself, $i \leftrightarrow i \leftrightarrow i$;
- $--if i \leftrightarrow ji \leftrightarrow j$, then $j \leftrightarrow ij \leftrightarrow i$;
- $--if i \leftrightarrow ji \leftrightarrow j \text{ and } j \leftrightarrow kj \leftrightarrow k, \text{ then } i \leftrightarrow ki \leftrightarrow k.$

Therefore, the states of a Markov chain can be partitioned into communicating *classes* such that only members of the same class communicate with each other. That is, two states ii and jj belong to the same class if and only if $i \leftrightarrow ji \leftrightarrow j$.

Example

Consider the Markov chain shown in Figure 11.9. It is assumed that when there is an arrow from state ii to state jj, then $p_{ij}>0$ pij>0. Find the equivalence classes for this Markov chain.



Solution

A Markov chain is said to be *irreducible* if it has only one communicating class. As we will see shortly, irreducibility is a desirable property in the sense that it can simplify analysis of the limiting behavior.

A Markov chain is said to be **irreducible** if all states communicate with each other.

Looking at Figure 11.10, we notice that there are two kinds of classes. In particular, if at any time the Markov chain enters Class 44, it will always stay in that class. On the other hand, for other classes this is not true. For example, if $X_0=1$ then the Markov chain might stay in Class 11 for a while, but at some point, it will leave that class and it will never return to that class again. The states in Class 44 are called *recurrent* states, while the other states in this chain are called *transient*.

In general, a state is said to be recurrent if, any time that we leave that state, we will return to that state in the future with probability one. On the other hand, if the probability of returning is less than one, the state is called transient. Here, we provide a formal definition:

For any state ii, we define

 $f_{ii}=P(X_n=i, \text{ for some } n\geq 1|X_0=i).f_{ii}=P(X_n=i, \text{ for some } n\geq 1|X_0=i).$

State ii is **recurrent** if fii=1fii=1, and it is **transient** if fii<1fii<1.

It is relatively easy to show that if two states are in the same class, either both of them are recurrent, or both of them are transient. Thus, we can extend the above definitions to classes. A class is said to be recurrent if the states in that class are recurrent. If, on the other hand, the states are transient, the class is called transient. In general, a Markov chain might consist of several transient classes as well as several recurrent classes.

Consider a Markov chain and assume $X_0=i \times 0=i$. If ii is a recurrent state, then the chain will return to state ii any time it leaves that state. Therefore, the chain will visit state ii an infinite number of times. On the other hand, if ii is a transient state, the chain will return to state ii with probability $f_{ii} < 1f_{ii} < 1$. Thus, in that case, the total number of visits to state ii will be a Geometric random variable with parameter $1-f_{ii}1-f_{ii}$.

Consider a discrete-time Markov chain. Let VV be the total number of visits to state ii.

a. If ii is a recurrent state, then

$$P(V=\infty|X_0=i)=1.P(V=\infty|X_0=i)=1.$$

b. If ii is a transient state, then

$$V|X_0=i\sim Geometric(1-fii).V|X_0=i\sim Geometric(1-fii).$$

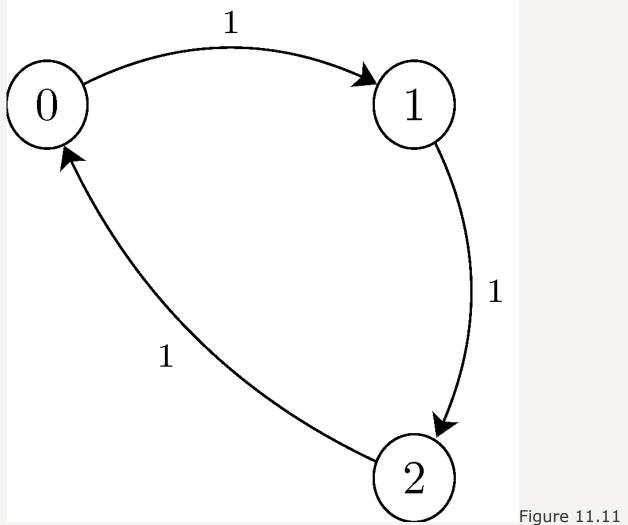
Example

Show that in a finite Markov chain, there is at least one recurrent class.

Solution

Periodicity:

Consider the Markov chain shown in Figure 11.11. There is a periodic pattern in this chain. Starting from state 00, we only return to 00 at times $n=3,6,\cdots n=3,6,\cdots$. In other words, p(n)00=0p00(n)=0, if nn is not divisible by 33. Such a state is called a *periodic* state with period d(0)=3d(0)=3.



- A state transition diagram.

The **period** of a state ii is the largest integer dd satisfying the following property: $p_{(n)ii}=0$ pii(n)=0, whenever nn is not divisible by dd. The period of ii is shown by d(i)d(i). If $p_{(n)ii}=0$ pii(n)=0, for all n>0n>0, then we let $d(i)=\infty$ d $(i)=\infty$.

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--If d(i)>1d(i)>1, we say that state ii is periodic.
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$$--$$
If $d(i)=1d(i)=1$, we say that state ii is **aperiodic**.

You can show that all states in the same communicating class have the same period. A class is said to be periodic if its states are periodic. Similarly, a class is said to be aperiodic if its states are aperiodic. Finally, a Markov chain is said to be aperiodic if all of its states are aperiodic.

If
$$i \leftrightarrow ji \leftrightarrow j$$
, then $d(i)=d(j)d(i)=d(j)$.

Why is periodicity important? As we will see shortly, it plays a roll when we discuss limiting distributions. It turns out that in a typical problem, we are given an irreducible Markov chain, and we need to check if it is aperiodic.

How do we check that a Markov chain is aperiodic? Here is a useful method. Remember that two numbers mm and ll are said to be *co-prime* if their greatest common divisor (gcd) is 11, i.e., gcd(l,m)=1gcd(l,m)=1. Now, suppose that we can find two co-prime numbers ll and mm such that p(l)ii>0pii(l)>0 and p(m)ii>0pii(m)>0. That is, we can go from state ii to itself in ll steps, and also in mm steps. Then, we can conclude state ii is aperiodic. If we have an irreducible Markov chain, this means that the chain is aperiodic. Since the number 11 is co-prime to every integer, any state with a self-transition is aperiodic.

Consider a finite irreducible Markov chain X_nX_n :

- a. If there is a self-transition in the chain $(p_{ii}>0p_{ii}>0$ for some ii), then the chain is aperiodic.
- b. Suppose that you can go from state ii to state ii in ll steps, i.e., p(l)ii>0pii(l)>0. Also suppose that p(m)ii>0pii(m)>0. If gcd(l,m)=1gcd(l,m)=1, then state ii is aperiodic.
- c. The chain is aperiodic if and only if there exists a positive integer nn such that all elements of the matrix P_nPn are strictly positive, i.e.,

p(n)ij > 0, for all $i,j \in S$.