

Classification of States of a Markov chain

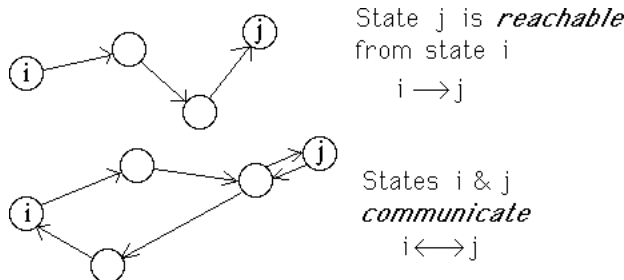
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A state i is *recurrent* if, given that the Markov chain starts in state i , the probability that it eventually returns to state i is one.

i.e.,
$$\sum_{n=1}^{\infty} f_{ii}^{(n)} = 1$$

$f_{ij}^{(n)}$ = Probability that the first visit to state j occurs at stage n , given that the initial state is i .

A state which is not recurrent is said to be *transient*.

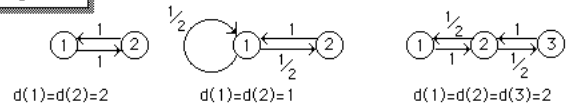


If state i is recurrent, and states i & j communicate, then state j is recurrent.

The *period* $d(i)$ of state i is the greatest common divisor of all the integers $n > 1$ for which

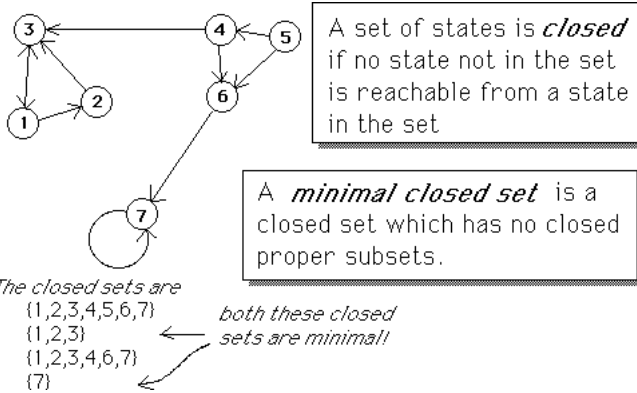
$$p_{ii}^{(n)} > 0$$

Examples



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

A Markov chain with $d(i)=1$ for all i is called *aperiodic*.



A minimal closed set is said to be *irreducible*.

A Markov chain is called *irreducible* if the set of its states is a minimal closed set.

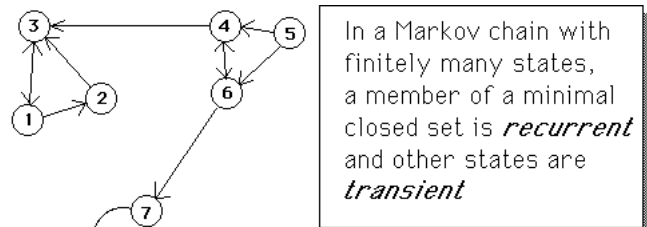
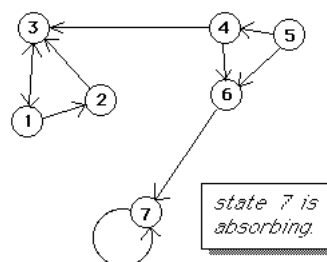
(A Markov chain is *irreducible* if and only if every pair of its states communicate.)

A state which forms a closed set, i.e., which cannot reach another state, is said to be *absorbing*.

If state j is absorbing, then

$$p_{jj}^{(n)} = 1$$

for all $n=1, 2, \dots$



States 1,2,3, & 7 are recurrent.

If state j is recurrent, but

$$\lim_{n \rightarrow \infty} p_{ij}^{(n)} = 0 \text{ for any state } i,$$

then state j is said to be *null*.

An irreducible Markov chain with *finitely* many states has

- no recurrent null states
- no transient states

The Powers of P

$$P = \begin{bmatrix} I & 0 \\ R & Q \end{bmatrix}$$

$$P^2 = \begin{bmatrix} I & 0 \\ R+QR & Q^2 \end{bmatrix}, \quad P^3 = \begin{bmatrix} I & 0 \\ R+QR+Q^2R & Q^3 \end{bmatrix}$$

$$\vdots$$

$$P^n = \begin{bmatrix} I & 0 \\ (I+Q+Q^2+\dots+Q^{n-1})R & Q^n \end{bmatrix} \left. \begin{array}{l} \text{absorbing} \\ \text{transient} \end{array} \right\}$$

Absorption Analysis

Consider a Markov chain with N states:

- r absorbing states
- $s = N-r$ transient states

Partition the transition probability matrix P :

$$P = \begin{bmatrix} I & 0 \\ R & Q \end{bmatrix} \left. \begin{array}{l} r \text{ rows} \\ s \text{ rows} \end{array} \right\}$$

$$\left. \begin{array}{l} r \\ s \end{array} \right\} \text{ columns}$$

Let states i & j both be transient, and define

e_{ij} = expected # of visits to state j , given that the system begins in state i (counting initial visit if $i=j$)

$$e_{ij} = \sum_{n=0}^{\infty} p_{ij}^{(n)}$$

and the $r \times r$ matrix:

$$E = \sum_{n=0}^{\infty} Q^n = (I - Q)^{-1}$$

since $(I - Q)(I + Q + Q^2 + \dots) = I + Q - Q + Q^2 - Q^2 + \dots = I$

Absorption probability

Let state i be transient and state j absorbing, and define:

a_{ij} = probability that the system enters the absorbing state j at some future time, given that it is initially in transient state i

$$a_{ij} = \sum_{n=1}^{\infty} f_{ij}^{(n)}$$

absorption probability (an infinite sum)

An alternate method for computing these probabilities:

Condition on the state entered at stage #1:

$$a_{ij} = \sum_{k=1}^N P(\text{system enters state } j | X_1 = k) P(X_1 = k)$$

$$= P(\text{system enters state } j | X_1 = j) P(X_1 = j)$$

$$+ \sum_{k \text{ absorbing, } \neq j} P(\text{system enters state } j | X_1 = k) P(X_1 = k)$$

$$+ \sum_{k \text{ transient}} P(\text{system enters state } j | X_1 = k) P(X_1 = k)$$

$$= 1p_{ij} + 0 + \sum_{k=1}^s a_{kj} p_{ik}$$

$$a_{ij} = p_{ij} + \sum_{k=1}^s a_{kj} p_{ik}$$

$$a_{ij} = p_{ij} + \sum_{k=1}^s a_{kj} p_{ik}, \text{ } i \text{ transient, } j \text{ absorbing}$$

In matrix form:

$$A = R + QA \text{ where } P = \begin{bmatrix} I & 0 \\ R & Q \end{bmatrix} \left. \begin{array}{l} \text{absorbing} \\ \text{transient} \end{array} \right\}$$

$$A - QA = R$$

$$(I - Q)A = R$$

$$A = (I - Q)^{-1}R$$

$$A = ER$$