

Review (Basics)

Def A Markov chain is a collection of random variables X_0, X_1, \dots over the same probability space with a common state space S such that

$$\mathbb{P}(X_{n+1} = i_{n+1} \mid X_0 = i_0, X_1 = i_1, \dots, X_{n-1} = i_{n-1}, X_n = i_n) = \mathbb{P}(X_{n+1} = i_{n+1} \mid X_n = i_n)$$

for any $i_0, i_1, \dots, i_{n+1} \in S$

In this class, a Markov chain is specified by:

① A finite state space $S: |S| < \infty$

② A time-homogeneous transition matrix \mathbb{P} :

$$\mathbb{P}(i, j) = \mathbb{P}(X_{n+1} = j \mid X_n = i) = \mathbb{P}(X_{n+1} = j \mid X_n = i) \text{ for any } n \geq 0.$$

③ An initial distribution π_0 :

$$\sum_{k \in S} \pi_0(k) = 1$$

Thm Let \mathbb{P}_n be the n -step transition matrix $\mathbb{P}_n(i, j) = \mathbb{P}(X_n = j \mid X_0 = i)$
 π_n be the distribution over the states after n steps.

Then

$$\mathbb{P}_n = \mathbb{P}^n$$
$$\pi_n = \pi_0 \mathbb{P}^n$$

Review (Conditioning)

Thm Let A, B, C, S be mutually exclusive. Let T_A, T_B be the times until we visit a state in A and B , respectively.

Define $\alpha(i) = P(T_A < T_B | X_0 = i)$, $\beta(i) = E[T_A | X_0 = i]$

Then

$$\alpha(i) = 1 \text{ if } i \in A$$

$$\alpha(i) = 0 \text{ if } i \in B$$

$$\begin{aligned} \alpha(i) &= \sum_{j \in S} P(T_A < T_B, X_1 = j | X_0 = i) \\ &= \sum_{j \in S} P(T_A < T_B | X_0 = i, X_1 = j) \cdot P(X_1 = j | X_0 = i) \\ &= \sum_{j \in S} P(T_A < T_B | X_0 = j) P(i, j) \text{ by homogeneity} \\ &= \sum_{j \in S} \alpha(j) P(i, j) \text{ if } i \notin A, i \notin B \end{aligned}$$

$$\beta(i) = 0 \text{ if } i \in A$$

$$\begin{aligned} \beta(i) &= \sum_{j \in S} E[T_A | X_0 = i, X_1 = j] \cdot P(X_1 = j | X_0 = i) \\ &= \sum_{j \in S} (1 + E[T_A | X_0 = j]) \cdot P(i, j) \text{ by homogeneity} \\ &= \sum_{j \in S} P(i, j) + \sum_{j \in S} \beta(j) P(i, j) \\ &= 1 + \sum_{j \in S} \beta(j) P(i, j) \text{ if } i \notin A \end{aligned}$$

Thm Let V_A be the number of times a state in A is visited.

Define $\delta(i) = E[V_A | X_0 = i]$

Then

$$\delta(i) = 1 + \sum_{j \in S} \delta(j) P(i, j) \text{ if } i \in A$$

$$\delta(i) = \sum_{j \in S} \delta(j) P(i, j) \text{ if } i \notin A$$

Note: $E[V_A]$ may be infinite

Compartmental Model

Ex Suppose that for some disease, the population can be split into three groups: susceptible, infected, and removed.

Each day,

- of the susceptible, 20% become infected; 80% stay susceptible.
- of the infected, 20% become susceptible; 10% become removed; 70% stay infected.
- of the removed, p become susceptible; $1-p$ stay removed.

In the population prior to exposure, 95% are susceptible and 5% are removed.

We can model this as a Markov process:

$$S = \{S, I, R\}$$

$$P = \begin{matrix} & \begin{matrix} S & I & R \end{matrix} \\ \begin{matrix} S \\ I \\ R \end{matrix} & \begin{bmatrix} 0.8 & 0.2 & 0 \\ 0.2 & 0.7 & 0.1 \\ p & 0 & 1-p \end{bmatrix} \end{matrix}$$

$$\pi_0 = [0.95 \quad 0 \quad 0.05]$$

[Demo]

Invariant Distributions

Def A distribution π is invariant for transition matrix P if $\pi = \pi P$

These equations are called the balance equations.

Thm $\pi_n = \pi_0$ for all $n \geq 0$ if and only if π_0 is invariant

Pf Suppose $\pi_n = \pi_0$ for all $n \geq 0$. Then

$$\pi_1 = \pi_0 P = \pi_0,$$

so π_0 is invariant.

Suppose $\pi_0 = \pi_0 P$. Then

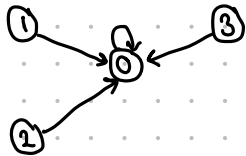
$$\pi_1 = \pi_0 P = \pi_0$$

$$\pi_{n+1} = \pi_n P = \pi_0$$

By induction, $\pi_n = \pi_0$ for all $n \geq 0$

Note A Markov chain may have many invariant distributions. For example, $P = I$ has infinitely many.

Note Invariance means that the net flow in and out of states is equal.



$\pi(0)$ of the particles are leaving state 0
 $\pi(0) \cdot P(0,0) + \pi(1) \cdot P(1,0) + \pi(2) \cdot P(2,0) + \pi(3) \cdot P(3,0)$ are entering state 0
The balance equations say these flows are equal:

$$\pi(i) = \underbrace{\sum_{j \in S} \pi(j) \cdot P(j,i)}_{\text{entering } i}$$

↑
leaving
i

Irreducibility

Def We say i can reach j ($i \rightarrow j$) if there is a path with positive probability from i to j .

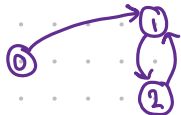
A Markov chain is irreducible if

$\forall i, j \quad i \rightarrow j$ and $j \rightarrow i$ (you can get from any state to any state)

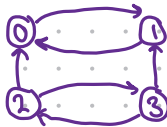
Ex Which of the following chains is irreducible?



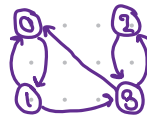
$0 \rightarrow 1$
Reducible



$1 \rightarrow 0$
Reducible



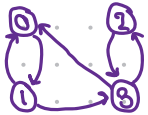
$0 \rightarrow 2$
Reducible



Irreducible

To show a chain is irreducible, construct a path that

- starts at any state
- goes through all of the other states
- ends at the starting state
- has positive probability



$0 \rightarrow 1 \rightarrow 3 \rightarrow 2 \rightarrow 3 \rightarrow 0$

Long Run Proportion of Time

Thm If a finite Markov chain is irreducible, then, for any initial distribution π_0 ,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{m=0}^{n-1} \mathbb{1}\{X_m = i\} = \pi(i) \quad \text{for all } i \in S$$

That is, the fraction of time spent in each state is given by π .

Also, π is invariant; therefore the invariant distribution exists and is unique.

Pf Because the chain is irreducible, each state will be visited infinitely many times. For each state i , let T_i be the number of steps to return to i starting with $X_0 = i$. Then let

$$\pi(i) = \frac{1}{\mathbb{E}[T_i]}, \quad \text{the fraction of time spent in state } i$$

So the fraction of time the chain spends in state i is $\pi(i)$, as desired.

Now, π is invariant. Note that in n steps, the chain is in state i for $n\pi(i)$ steps.

Over large n , the chain transitions from any state j to i

$$n\pi(j) \cdot \underbrace{P(j, i)}_{\substack{\text{in state } j \\ \text{transition} \\ \text{to state } i}}$$

times. The total visits to state i is the sum of all visits from the states:

$$n\pi(i) = \sum_{j \in S} n\pi(j)P(j, i)$$

This can be written as $\pi = \pi P$.

Proof of uniqueness is more complicated.

Note An irreducible Markov chain's distribution does not necessarily converge to π :



The chain spends half the time in each state, but every initial distribution does not converge to $[\frac{1}{2}, \frac{1}{2}]$, e.g.

$$[0 \ 1] \rightarrow [1 \ 0] \rightarrow [0 \ 1] \rightarrow \dots$$

This is because of the periodic behavior of the chain.

Periodicity

Def The period of a state i , denoted $d(i)$, is

$$d(i) = \gcd \{ n : P_n(i, i) > 0 \};$$

that is, the period of a state is the greatest common divisor of the lengths of paths from i to i .

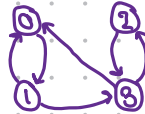
In this class we only consider the period of states in irreducible chains.

Ex Find $d(i)$ for each of the following chains



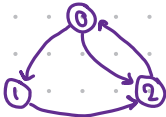
$1 \rightarrow 2 \rightarrow 1; 1 \rightarrow 2 \rightarrow 1 \rightarrow 2 \rightarrow 1$

$$d(i) = \gcd \{ 2, 4, 6, \dots \} = 2$$



$1 \rightarrow 0 \rightarrow 1; 1 \rightarrow 3 \rightarrow 0 \rightarrow 1;$

$$d(i) = \gcd \{ 3, 3, \dots \} = 1$$



$1 \rightarrow 2 \rightarrow 0 \rightarrow 1; 1 \rightarrow 2 \rightarrow 0 \rightarrow 2 \rightarrow 1$

$$d(i) = \gcd \{ 3, 4, \dots \} = 1$$

Clm If a Markov chain is irreducible, the periods of all states are the same:

$$d(i) = d(j) \text{ for all } i, j \in S$$

Pf As an exercise

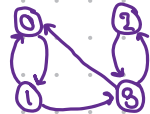
Def If $d(i) = 1$ for any state in an irreducible Markov chain, we say the chain is aperiodic

Ex Which of the following chains is aperiodic?



periodic, since

$$d(i) = 2$$



aperiodic, since

$$d(i) = 1$$



aperiodic, since

$$d(i) = 1$$

To show a chain is aperiodic, find two loops with coprime lengths

Markov Chain Convergence Theorem

Thm (Finite Markov Chain Convergence) If X_0, X_1, \dots is a Markov chain on S with time-homogeneous transition matrix P and

① $|S| < \infty$

② P is irreducible: $\forall i, j \in S, \exists n P_n(i, j) > 0$

③ P is aperiodic: $\forall i \in S, d(i) = 1$,

then X_0, X_1, \dots has an invariant distribution $\pi = \pi P$ such that

$$\lim_{n \rightarrow \infty} P_n(i, j) = \pi(j)$$

i.e. the n -step transition probabilities converge to π .

Moreover,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} \mathbb{1}_{\{X_k = j\}} = \pi(j),$$

i.e. $\pi(j)$ is the long-run proportion of time spent in state j .

Q
P

$$\pi_n(i) = P(X_n = i) \rightarrow \pi(i) \text{ as } n \rightarrow \infty$$

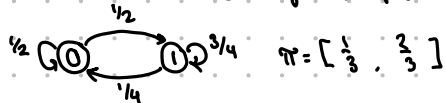
$$\pi_n(i) = \sum_{j \in S} \pi_0(j) P_n(j, i)$$

$$\rightarrow \sum_{j \in S} \pi_0(j) \pi(i) \text{ as } n \rightarrow \infty$$

$$= \pi(i) \sum_{j \in S} \pi_0(j)$$

$$= \pi(i)$$

Note We can find the long-run probability of an event by conditioning on π .



What is the long-run probability that we stay in the same state?

$$\begin{aligned} P(\text{stay}) &= P(\text{stay} \mid \text{state } 0) \cdot P(\text{state } 0) + P(\text{stay} \mid \text{state } 1) \cdot P(\text{state } 1) \\ &= \frac{1}{2} \cdot \frac{1}{3} + \frac{3}{4} \cdot \frac{2}{3} \\ &= \frac{5}{6} \end{aligned}$$

Ehrenfest Chain 1

Q In the Ehrenfest model, there are two containers, containing a total of N particles.

At each step

- a container is selected uniformly at random
 - a particle is selected uniformly at random, independently of the container
- and the selected particle is placed in the selected container; if the particle was already in the container, it remains in place

Let X_n be the number of particles in the first container at time n .

What are the transition probabilities of the chain?

The number of particles can either increase by 1, decrease by 1, or remain the same.

$$P(X_{n+1} = i+1 \mid X_n = i) = \frac{1}{2} \cdot \frac{N-i}{N} = \frac{N-i}{2N}$$

$$P(X_{n+1} = i-1 \mid X_n = i-1) = \frac{1}{2} \cdot \frac{i}{N} = \frac{i}{2N}$$

$$P(X_{n+1} = i \mid X_n = i) = 1 - \frac{i}{2N} - \frac{N-i}{2N} = \frac{1}{2}$$

So

$$P(i, j) = \begin{cases} \frac{N-i}{2N} & \text{if } j = i+1 \\ \frac{1}{2} & \text{if } j = i \\ \frac{i}{2N} & \text{if } j = i-1 \\ 0 & \text{otherwise} \end{cases}$$

Prove that any distribution over the states converges to some distribution π .

We must show that the chain is irreducible and aperiodic

Irreducible: consider the path $0 \rightarrow 1 \rightarrow 2 \rightarrow \dots \rightarrow N \rightarrow N-1 \rightarrow \dots \rightarrow 0$.

Aperiodic: note that $d(0) = \gcd\{1, \dots\} = 1$. So all states have period 1.

Ehrenfest Chain II

Q In the Ehrenfest model, there are two containers, containing a total of N particles.

At each step

- a container is selected uniformly at random
- a particle is selected uniformly at random, independently of the container

and the selected particle is placed in the selected container; if the particle was already in the container, it remains in place.

Let X_n be the number of particles in the first container at time n .

Find the stationary distribution of the chain.

The balance equations are

$$\pi(0) = \frac{1}{2} \pi(0) + \frac{1}{2N} \pi(1)$$

$$\pi(j) = \frac{N-(j-1)}{2N} \pi(j-1) + \frac{1}{2} \pi(j) + \frac{j+1}{2N} \pi(j+1) \quad \text{for } 1 \leq j \leq N-1$$

$$\pi(N) = \frac{1}{2N} \pi(N-1) + \frac{1}{2} \pi(N)$$

Rewrite the first few equations in terms of $\pi(0)$

$$\pi(0) = \frac{1}{2} \pi(0) + \frac{1}{2N} \pi(1) \Rightarrow \pi(1) = N \pi(0) = \binom{N}{1} \pi(0)$$

$$\pi(1) = \frac{N}{2N} \pi(0) + \frac{1}{2} \pi(1) + \frac{2}{2N} \pi(2) \Rightarrow \pi(2) = \frac{N}{2} (\pi(1) - \pi(0)) = \frac{N(N-1)}{2} \pi(0) = \binom{N}{2} \pi(0)$$

By induction, $\pi(j) = \binom{N}{j} \pi(0)$.

Then

$$\sum_{j=0}^N \pi(j) = \sum_{j=0}^N \binom{N}{j} \pi(0) = \pi(0) \sum_{j=0}^N \binom{N}{j} = \pi(0) 2^N = 1 \Rightarrow \pi(0) = \frac{1}{2^N}$$

So

$$\pi(k) = P(X_n = k) = \binom{N}{k} \frac{1}{2^N}$$

i.e. $\pi(k) \sim \text{Binomial}(N, \frac{1}{2})$.