

Markov Chain — Core Concepts (Step by Step)

What is a Markov Chain? (Quick reminder)

A **Markov chain** is a stochastic process

X_0, X_1, X_2, \dots

where the **future depends only on the present**, not the past:

$$P(X_{n+1}=j | X_n=i, X_{n-1}, \dots) = P(X_{n+1}=j | X_n=i) \\ P(X_{n+1}=j | X_n=i, X_{n-1}, \dots) = P(X_{n+1}=j | X_n=i)$$

This is called the **Markov property**.

1 What is the Transition Matrix? [2p]

Definition

The **transition matrix** P contains the probabilities of moving from one state to another in **one step**.

If the state space is $S = \{1, 2, \dots, k\}$, then

$$P = (p_{ij}), \text{ where } p_{ij} = P(X_{n+1}=j | X_n=i) \\ P = (p_{ij}), \quad \text{where } p_{ij} = P(X_{n+1}=j | X_n=i)$$

Properties

- $p_{ij} \geq 0$
- Each row sums to 1:

$$\sum_j p_{ij} = 1$$

Example

Suppose we have **3 states**:

- State 1
- State 2
- State 3

Transition probabilities:

- From state 1 \rightarrow 2 with prob 1
- From state 2 \rightarrow 1 or 3 (each 0.5)
- From state 3 \rightarrow 3 with prob 1

Then

$$P = \begin{pmatrix} 0 & 1 & 0 \\ 0.5 & 0 & 0.5 \\ 0 & 0 & 1 \end{pmatrix}$$

Intuition

Each **row** answers:

“If I am in state i now, where can I be next?”

2 Is the Markov Chain Irreducible? [2p]

Definition

A Markov chain is **irreducible** if **every state can be reached from every other state** (possibly in multiple steps).

Mathematically:

$$\forall i, j \exists n \geq 1 \text{ such that } P^n(i, j) > 0 \quad \text{for all } i, j \quad \Leftrightarrow \quad \exists n \geq 1 \text{ such that } P^n(i, j) > 0 \quad \forall i, j \exists n \geq 1 \text{ such that } P^n(i, j) > 0$$

Example 1: Irreducible Chain

$$P = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

- From 1 \rightarrow 2
- From 2 \rightarrow 1

✓ Every state communicates with every other \rightarrow **irreducible**

Example 2: Not Irreducible

$$P = \begin{pmatrix} 1 & 0 \\ 0.4 & 0.6 \end{pmatrix}$$

- From state 1 \rightarrow stuck forever
- State 2 can reach 1, but 1 cannot reach 2

✗ **Not irreducible**

Intuition

Can I eventually go **everywhere** from **anywhere**?

3 Is the Markov Chain Aperiodic? What is the Period? [3p]

Definition of Period

The **period** of state i is:

$$d(i) = \gcd\{n \geq 1 : P^n(i, i) > 0\}$$

- If $d(i) = 1$, state is **aperiodic**
 - If $d(i) > 1$, state is **periodic**
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Example 1: Periodic Chain

$$P = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

- From state 1 \rightarrow back to 1 only in **even steps**
- Possible return times: 2, 4, 6, ...

$$d(1) = \gcd(2, 4, 6, \dots) = 2$$

✗ Periodic with period 2

Example 2: Aperiodic Chain

$$P = \begin{pmatrix} 0.5 & 0.5 \\ 0.3 & 0.7 \end{pmatrix}$$

- You can stay in the same state in **1 step**
- So return time includes 1

$$d(i) = 1$$

✓ Aperiodic

Important Theorem (Exam Favorite)

If a Markov chain is **irreducible**, then **all states have the same period**.

Intuition

Are you forced to return in a **fixed cycle**, or can you come back at **any time**?

4 Does the Markov Chain Have a Stationary Distribution? [3p]

Definition

A **stationary distribution** π satisfies:

$$\pi = \pi P \text{ and } \sum_i \pi_i = 1 \quad \pi_i = \sum_j \pi_j P_{ji} \quad \sum_i \pi_i = 1$$

This means:

If the chain starts with distribution π , it **stays the same forever**.

Existence Condition (Key Result)

A **finite, irreducible** Markov chain **always has** a stationary distribution.

Example

$$P = \begin{pmatrix} 0.7 & 0.3 \\ 0.4 & 0.6 \end{pmatrix}$$

$$\pi = (\pi_1, \pi_2)$$

Solve:

$$\begin{cases} \pi_1 = 0.7\pi_1 + 0.4\pi_2 \\ \pi_2 = 0.3\pi_1 + 0.6\pi_2 \\ \pi_1 + \pi_2 = 1 \end{cases}$$

From first equation:

$$0.3\pi_1 = 0.4\pi_2 \Rightarrow \pi_1 = \frac{4}{3}\pi_2$$

Normalization:

$$\frac{4}{3}\pi_2 + \pi_2 = 1 \Rightarrow \pi_2 = \frac{3}{7}, \pi_1 = \frac{4}{7}$$

✓ Stationary distribution:

$$\pi = \left(\frac{4}{7}, \frac{3}{7} \right)$$

Intuition

Long-run proportion of time spent in each state

5 Is the Markov Chain Reversible? [3p]

Definition

A Markov chain is **reversible** if there exists a stationary distribution π such that:

$$\pi_i p_{ij} = \pi_j p_{ji} \quad \forall i, j \quad \pi_i p_{i|} p_{ij} = \pi_j p_{ji} \quad \forall i, j$$

This is called **detailed balance**.

Physical Meaning

Flow from $i \rightarrow j$ equals flow from $j \rightarrow i$

Example: Reversible Chain

Using the previous example:

$$\pi = (47, 37) \quad \pi = \left(\frac{4}{7}, \frac{3}{7} \right) \quad \pi = (74, 73)$$

Check:

$$\pi_1 p_{12} = 47 \times 0.3 = 12.7 \quad \pi_1 p_{12} = \frac{4}{7} \times 0.3 = \frac{12}{70} \quad \pi_1 p_{12} = 74 \times 0.3 = 22.2$$

$$\pi_2 p_{21} = 37 \times 0.4 = 14.8 \quad \pi_2 p_{21} = \frac{3}{7} \times 0.4 = \frac{12}{70} \quad \pi_2 p_{21} = 73 \times 0.4 = 29.2$$

✓ Detailed balance holds \rightarrow **reversible**

Example: Non-Reversible Chain

If

$$\pi_i p_{ij} \neq \pi_j p_{ji} \quad \pi_i p_{ij} \neq \pi_j p_{ji}$$

for **any pair**, chain is **not reversible**.

Intuition

Running the movie backward looks statistically the same 🎬

🔑 Summary Table (Exam Gold)

Concept	Key Idea
Transition Matrix	One-step transition probabilities
Irreducible	All states communicate
Aperiodic	No fixed return cycle
Stationary Distribution	Long-run stable distribution
Reversible	Detailed balance holds

If you want, next I can:

- Solve a **full exam-style problem**
- Explain **convergence to stationarity**
- Connect this with **MCMC / Metropolis–Hastings**
- Give **intuition using graphs**

Just tell me 👍