

# Markov Chain — Core Concepts (Step by Step)

## What is a Markov Chain? (Quick reminder)

A **Markov chain** is a stochastic process

$X_0, X_1, X_2, \dots, X_{-0}, X_{-1}, X_{-2}, \dots, X_0, X_1, X_2, \dots$

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

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

This is called the **Markov property**.

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## 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

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

### Properties

- $p_{ij} \geq 0$  and  $\sum_j p_{ij} = 1$
- 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 → 2 with prob 1
- From state 2 → 1 or 3 (each 0.5)
- From state 3 → 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?”

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## 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:

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

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## Example 1: Irreducible Chain

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

- From 1 → 2
- From 2 → 1

✓ Every state communicates with every other → **irreducible**

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## Example 2: Not Irreducible

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

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

✗ Not irreducible

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## Intuition

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

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

### Definition of Period

The **period** of state  $i$  is:

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

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

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

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

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

 **Periodic with period 2**

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## Example 2: Aperiodic Chain

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

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

$$d(i)=1 \\ d(i) = 1 \\ d(i)=1$$

 **Aperiodic**

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## Important Theorem (Exam Favorite)

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

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## Intuition

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

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

### Definition

A **stationary distribution**  $\pi$  satisfies:

$$\pi = \pi P \text{ and } \sum_i \pi_i = 1 \quad \text{and} \quad \sum_i \pi_i = 1$$

This means:

If the chain starts with distribution  $\pi$ , it **stays the same forever**.

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## Existence Condition (Key Result)

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

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## Example

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

Let  $\pi = (\pi_1, \pi_2)$  where  $\pi_1 = \pi_1, \pi_2 = \pi_2$ .

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:

$$0.3\pi_1 = 0.4\pi_2 \Rightarrow \pi_1 = \frac{4}{3}\pi_2 \Rightarrow \pi_1 = \frac{4}{7}\pi_2$$

Normalization:

$$\begin{aligned} \pi_1 + \pi_2 = 1 &\Rightarrow \frac{4}{7}\pi_2 + \pi_2 = 1 \Rightarrow \pi_2 = \frac{7}{11} \\ \pi_1 &= \frac{4}{7}\pi_2 = \frac{4}{7} \cdot \frac{7}{11} = \frac{4}{11} \end{aligned}$$

✓ Stationary distribution:

$$\pi = \left( \frac{4}{11}, \frac{7}{11} \right)$$

## Intuition

Long-run proportion of time spent in each state

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## 5 Is the Markov Chain Reversible? [3p]

### Definition

A Markov chain is **reversible** if there exists a stationary distribution  $\pi$  such that:

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

This is called **detailed balance**.

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### Physical Meaning

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

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### Example: Reversible Chain

Using the previous example:

$$\pi = (47, 37) \text{ where } \pi_1 = \frac{4}{7}, \pi_2 = \frac{3}{7}$$

Check:

$$\pi_1 p_{12} = 47 \times 0.3 = 1270 \quad \pi_1 p_{12} = \frac{4}{7} \times 0.3 = \frac{12}{70}$$
$$\pi_2 p_{21} = 37 \times 0.4 = 1270 \quad \pi_2 p_{21} = \frac{3}{7} \times 0.4 = \frac{12}{70}$$

✓ Detailed balance holds  $\rightarrow$  **reversible**

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### Example: Non-Reversible Chain

If

$$\pi_i p_{ij} \neq \pi_j p_{ji} \quad \text{for any pair } i, j$$

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

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## Intuition

Running the movie backward looks statistically the same 🎥

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## 🔑 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

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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 👍