

# Markov Chains

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# What is a Markov Chain?

**Definition:** A discrete-time process *Markov Chain* is a sequence of random variables  $\{X_t\}_{t \geq 0}$  with the property that, given the present state, the future and past states are independent. Formally,

$$P(X_{t+1}|X_t, X_{t-1}, \dots, X_0) = P(X_{t+1}|X_t).$$

The Markov Chain is **time-homogeneous** if the transition probabilities do not depend on time  $t$ :

$$\forall n \in \mathbb{N}, \quad P(X_t = y | X_{t-1} = x) = P(X_{t+m} = y | X_{t+m-1} = x)$$

i.e. transition probabilities do not depend on  $t$ . The **key idea** is MCMC is to construct a Markov Chains such that  $x_t$  converges to a desired distribution  $\pi$  as  $t \rightarrow \infty$  and

$$\frac{1}{n} \sum_{t=1}^n \phi(x_t) \rightarrow \mathbb{E}_{x \sim \pi}[\phi(x)] \quad \text{as } n \rightarrow \infty.$$

What kinds of conditions are required for this to hold?

# Invariant / stationary distribution

A distribution  $\pi$  is called **invariant** (or **stationary**) for a Markov Chain with transition kernel  $P$  if

$$\pi(y) = \int \pi(x)P(x, y)dx.$$

Intuitively, if the chain starts with distribution  $\pi$ , it remains in distribution  $\pi$  at all future times.

Time-homogeneous is not needed for invariant distribution. But it is often easier to verify in that case.

# Irreducible

A Markov Chain is called **irreducible** if it is possible to get to any state from any state. Formally, for any states  $x$  and  $y$ , there exists an integer  $0 \leq n < \infty$  such that

$$P^n(x, y) > 0,$$

where  $P^n(x, y)$  is the  $n$ -step transition probability from state  $x$  to state  $y$ .

# Aperiodicity

A Markov Chain is called **aperiodic** if it does not get trapped in cycles with fixed periods. Ensures actual convergence instead of oscillation. Formally, for any state  $x$ , the greatest common divisor of the set of integers

$$\{n \geq 1 : P^n(x, x) > 0\}$$

is 1. **Note:** If all states have a non-zero probability of remaining in the same state, the chain is aperiodic.

# Positive recurrence

A Markov Chain is called **positive recurrent** if, starting from any state, the expected return time to that state is finite. Formally, for any state  $x$ ,

$$\mathbb{E}[T_x | X_0 = x] < \infty,$$

where  $T_x$  is the return time to state  $x$ . **Note:** Positive recurrence ensures that the chain does not wander off to infinity and has a well-defined long-term behavior.

# More on recurrence

- ▶ **Recurrent:** A Markov Chain is called recurrent if, starting from any state, the probability of returning to that state is 1.
- ▶ **Positive recurrence:** A Markov Chain is called positive recurrent if it is recurrent and the expected return time to any state is finite, i.e. the chain returns quickly on average.
- ▶ **Transient:**
- ▶ **Null recurrence:** A Markov Chain is called null recurrent if it is recurrent but the expected return time to any state is infinite.