

# 18.445 Introduction to Stochastic Processes

## Lecture 3: Markov chains: time-reversal

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## Recall

Consider a Markov chain with state space  $\Omega$  and transition matrix  $P$  :

$$\mathbb{P}[X_{n+1} = y \mid X_n = x] = P(x, y).$$

- A probability measure  $\pi$  is stationary if  $\pi = \pi P$ .
- If  $P$  is irreducible, there exists a unique stationary distribution.

## Today's goal

- Ergodic Theorem
- Time-reversal of Markov chain
- Birth-and-Death chains
- Total variation distance

# Ergodic Theorem

## Theorem

Let  $f$  be a real-valued function defined on  $\Omega$ . If  $(X_n)_n$  is an irreducible Markov chain with stationary distribution  $\pi$ , then for any starting distribution  $\mu$ , we have

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^n f(X_j) = \pi f, \quad \mathbb{P}_\mu - \text{a.s.}$$

In particular,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^n 1_{[X_j=x]} = \pi(x), \quad \mathbb{P}_\mu - \text{a.s.}$$

# Detailed balance equations

## Definition

Suppose that a probability measure  $\pi$  on  $\Omega$  satisfies

$$\pi(x)P(x, y) = \pi(y)P(y, x), \quad \forall x, y \in \Omega.$$

These are called detailed balance equations.

## Lemma

*Any distribution  $\pi$  satisfying the detailed balance equations is stationary for  $P$ .*

## Definition

A chain satisfying detailed balance equations is called reversible.

# Simple random walk on graph

**Example** Consider simple random walk on graph  $G = (V, E)$  (which is connected). The measure

$$\pi(x) = \frac{\text{deg}(x)}{2|E|}, \quad x \in \Omega$$

satisfies detailed balance equations ; therefore the simple random walk on  $G$  is reversible.

# Time-reversal of Markov chain

## Theorem

Let  $(X_n)$  be an irreducible Markov chain with transition matrix  $P$  and stationary distribution  $\pi$ . Define  $\hat{P}$  to be

$$\hat{P}(x, y) = \frac{\pi(y)P(y, x)}{\pi(x)}.$$

- $\hat{P}$  is stochastic
- Let  $(\hat{X}_n)$  be a Markov chain with transition matrix  $\hat{P}$ . Then  $\pi$  is also stationary for  $\hat{P}$ .
- For any  $x_0, \dots, x_n \in \Omega$ , we have

$$\mathbb{P}_\pi[X_0 = x_0, \dots, X_n = x_n] = \mathbb{P}_\pi[\hat{X}_0 = x_n, \dots, \hat{X}_n = x_0].$$

We call  $\hat{X}$  the time-reversal of  $X$ .

**Remark** If a chain with transition matrix  $P$  is reversible, then  $\hat{P} = P$  and  $\hat{X}$  has the same law as  $X$ .

# Birth-and-Death chains

A birth-and-death chain has state space  $\Omega = \{0, 1, \dots, N\}$ .

The current state can be thought of as the size of some population; in a single step of the chain there can be at most one birth or death. The transition probabilities can be specified by  $\{(p_k, r_k, q_k)_{k=0}^N\}$  where  $p_k + r_k + q_k = 1$  for each  $k$  and

- $p_k$  is the probability of moving from  $k$  to  $k + 1$  when  $0 \leq k < N$ ;  
 $p_N = 0$
- $q_k$  is the probability of moving from  $k$  to  $k - 1$  when  $0 < k \leq N$ ;  
 $q_0 = 0$
- $r_k$  is the probability of remaining at  $k$  when  $0 \leq k \leq N$ .

## Theorem

*Every birth-and-death chain is reversible.*

# Total variation distance

## Definition

The total variation distance between two probability measures  $\mu$  and  $\nu$  on  $\Omega$  is defined by

$$\|\mu - \nu\|_{TV} = \max_{A \subset \Omega} |\mu(A) - \nu(A)|.$$

## Lemma

*The total variation distance satisfies triangle inequality :*

$$\|\mu - \nu\|_{TV} \leq \|\mu - \eta\|_{TV} + \|\eta - \nu\|_{TV}.$$



# Three ways to characterize the total variation distance

Lemma

$$\|\mu - \nu\|_{TV} = \frac{1}{2} \sum_{x \in \Omega} |\mu(x) - \nu(x)|.$$

Lemma

$$\|\mu - \nu\|_{TV} = \frac{1}{2} \sup\{\mu f - \nu f : f \text{ satisfying } \max_{x \in \Omega} |f(x)| \leq 1\}.$$

# Three ways to characterize the total variation distance

## Definition

A coupling of two probability measures  $\mu$  and  $\nu$  is a pair of random variables  $(X, Y)$  defined on the same probability space such that the marginal law of  $X$  is  $\mu$  and the marginal law of  $Y$  is  $\nu$ .

## Lemma

$$\|\mu - \nu\|_{TV} = \inf\{\mathbb{P}[X \neq Y] : (X, Y) \text{ is a coupling of } \mu, \nu\}.$$

## Definition

We call  $(X, Y)$  the optimal coupling if  $\mathbb{P}[X \neq Y] = \|\mu - \nu\|_{TV}$ .

# Homework 1 due Feb. 23rd

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