# Ergodicity of Markov processes: theory and computation

#### Yao Li

Department of Mathematics and Statistics, University of Massachusetts Amherst

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ICERM, Brown University



#### Outline

- Markov processes on measurable state space.
- Coupling method and renewal theory
- Exponential and power-law ergodicity
- Construction of Lyapunov functions
- Numerical computation of ergodicity
- Numerical computation of invariant probability measures

# Basic setting 1

- $\bullet$   $\Phi_n$  discrete time Markov process
- $(X, \mathcal{B}(X))$  state space with a sigma algebra  $\mathcal{B}(X)$
- **③** P transition probability.  $P(x, A) = \mathbb{P}[\Phi_1 \in A \mid \Phi_0 = x]$ .
- **②**  $P(x, \cdot)$  is a probability measure on  $(X, \mathcal{B}(X))$ , P(x, A) is a measurable function for any  $A \in \mathcal{B}$ .
- By Markov property, this is enough to determine a Markov process

# Basic setting 2

Markov property: only depends on the nearest history

$$\mathbb{P}[\Phi_{n+1} \in A \,|\, \Phi_0, \cdots, \Phi_n] = \mathbb{P}[\Phi_{n+1} \in A \,|\, \Phi_n]$$

 $P^m(x,A) = \mathbb{P}[\Phi_{n+m} \in A \mid \Phi_n = x].$ 

•

$$P^{m+n}(x,A) = \int_X P^n(y,A) P^m(x,dy)$$

- First arrival time:  $\eta_A = \inf_{n \geq 1} \{ \Phi_n \in A \}$
- Note that  $\eta_A$  is a stopping time (random time that only depends on historical and present states of  $\Phi_n$ .)
- Hitting probability:  $L(x, A) = \mathbb{P}[\Phi_n \in A \text{ for some } n | \Phi_0 = x]$

# Irreducibility

Main difference from discrete Markov chain: P(x, y) does not make sense any more!

 $\Phi_n$  is irreducible if there exists a reference measure  $\psi$  on X such that

- If  $\psi(A) > 0$ , then L(x, A) > 0 for all  $x \in X$
- ② If  $\psi(A) = 0$ , then  $\psi(\{y : L(y, A) > 0\}) = 0$

 $\Phi_n$  can reach everywhere that could be "seen" by  $\psi$ .

# Example

Stochastic differential equation  $X_t$ . Euler-Maruyama method.

$$X_{n+1} = X_n + f(X_n)h + \sigma(X_n)\mathcal{N}(0,1)\sqrt{h}$$

Transition kernel

$$P(x,A) = \int_A \frac{1}{\sqrt{2\pi\sigma(x)^2 h}} e^{-(y-x-f(x)h)^2/2\sigma^2(x)h} dy$$

Let Lebesgue measure be the reference measure. Easy to check that  $X_n$  is irreducible.

### Atom and pseudo-atom

- ① Discrete state space: P(x, y) > 0. Very useful!
- **②** Atom:  $\alpha$  is an atom if  $P(x, \cdot) = P(y, \cdot)$  for all  $x, y \in \alpha$ . Atom is like a discrete state.
- Atom usually does not exist
- Pseudo-atom: small set C
- **⊙**  $C \in \mathcal{B}(X)$  is a small set if there exist an integer  $n \in \mathbb{N}$  and a nontrivial measure  $\nu$  such that

$$P^n(x,A) \ge \nu(A)$$
 for all  $x \in C$ 



# Example

Euler-Maruyama scheme again

$$X_{n+1} = X_n + f(X_n)h + \sigma(X_n)\mathcal{N}(0,1)\sqrt{h}$$

Every bounded set is a small set because the probability density of P is everywhere strictly positive.

Random walk:  $X_{n+1}=X_n+U_n$ ,  $U_n\sim U(-1/2,1/2)$ .  $[-1/4\,,\,1/4]$  is a small set with n=1 and  $\nu=$  Lebesgue measure.

# (A)periodicity

#### Discrete space

Assume irreducibility. Define  $E = \{n \mid P^n(x, x) > 0\}$ . Period d is the greatest common divisor of E.

#### General space

Assume irreducibility. C is a small set. Define

$$E_C = \{ n \mid P^n(x, \cdot) \ge \nu(\cdot), x \in C, \nu(C) > 0 \}$$

(positive probability that the chain will return to C after n steps.) Period d is the greatest common divisor of E.

 $\Phi_n$  is aperiodic if d=1.



# **Ergodicity**

From now on we assume that  $\Phi_n$  is irreducible and aperiodic.

- Left operator:  $\mu$  probability measure.  $\mu P^n(A) = \mathbb{P}_{\mu}[\Phi_n \in A]$ .
- **2** Right operator: f observable (function).  $P^n f(x) = \mathbb{E}_x[f(\Phi_n)]$ .
- **1** Invariant probability measure.  $\pi$  is said to be invariant if  $\pi P = \pi$ .

Let  $\mu$  and  $\nu$  be two probability measures. Does

$$\|\mu P^n - \nu P^n\|_{TV}$$

converge to zero? If yes, how fast??



# Main approach: Coupling

A Markov process  $(\Phi_n^1, \Phi_n^2)$  on the state space  $X \times X$  is said to be a Markov coupling if

- Two marginal distributions are Markov processes  $\Phi_n$  with initial distribution  $\mu$  and  $\nu$ , respectively
- ② If  $\Phi_n^1 = \Phi_n^2$ , then  $\Phi_m^1 = \Phi_m^2$  for all  $m \ge n$ .

 $\tau_C = \inf_{n \geq 0} \{ \Phi_n^1 = \Phi_n^2 \}$  is the *coupling time*.

# Coupling Lemma

#### Coupling Lemma

$$\|\mu P^n - \nu P^n\|_{TV} \le 2\mathbb{P}[\tau_C > n].$$

(See whiteboard for the proof. )

#### Optimal coupling (Pitman 1970s)

There exists a coupling  $(\Phi_n^1, \Phi_n^2)$  (may not be Markov) such that

$$\|\mu P^n - \nu P^n\|_{TV} = 2\mathbb{P}[\tau_C > n].$$

The existence of "honest" optimal coupling remains open.

# Coupling at atom

- **1** Assume  $\Phi_n$  admits an atom  $\alpha$ .
- ② Let  $(\Phi_n^1, \Phi_n^2)$  be a coupling such that  $\Phi_n^1$  and  $\Phi_n^2$  are independent until their first simultaneous visit to  $\alpha$ , and run together after that.

Easy to check:  $(\Phi_n^1, \Phi_n^2)$  is a Markov coupling.

Difficulty: property of  $\mathbb{P}[\tau_C > n]$ ?

- ② Exponential:  $\mathbb{P}[\tau_C > n] \sim \rho^{-n}$  for  $\rho > 1$
- ② Power-law:  $\mathbb{P}[\tau_C > n] \sim n^{-\beta}$  for  $\beta > 0$

# Renewal process

Let

$$S_n = \sum_{i=0}^n Y_i$$

such that  $Y_1, Y_2, \cdots$  are i.i.d. random nonnegative integers. ( $Y_0$  could be different).  $S_n$  is a renewal process.  $Y_i$  is called inter-occurrence time.

Let  $u_n = \mathbb{P}[n = S_m \text{ for some m}].$ If S is aperiodic,  $u_n \to 1/\mathbb{E}[Y_1].$ 

# Renewal process from $\Phi_n$

- $\mathbf{Q}$   $\alpha$  is the atom.
- $Y_0 = \eta_\alpha$
- **3**  $S_n$  is the *n*-th visit to  $\alpha$

#### Simultaneous renewal

- Now let  $S_n$  and  $S'_n$  be two renewal processes corresponding to  $\Phi^1_n$  and  $\Phi^2_n$ , respectively.
- ② The coupling time  $\tau_C$  is the first simultaneous renewal time.

$$\tau_C = \inf_n \{ n = S_{k_1} = S'_{k_2} \text{ for some } k_1 \text{ and } k_2 \}$$

#### Three questions

- 1 What if there is no atom? ✓
- 2 First simultaneous renewal time? ✓
- 3 How to estimate the first visit time  $\eta_{\alpha}$  (probably tomorrow)

# How to make an atom? (1)

- Atom does not exist in most scenarios
- Small set is much easier to get
- $\odot$  Simplest case. Let C be a small set that satisfies

$$P(x, A) \ge \delta \mathbf{1}_{C}(x)\nu(A)$$
 ,  $A \in \mathcal{B}(X), x \in X$ ,

where  $\nu$  is a probability measure with  $\nu(C) = 1$ .

- Split X into  $\hat{X} = X \times \{0,1\}$  with  $X_0 = X \times \{0\}$  and  $X_1 = X \times \{1\}$ .
- $\odot$  Similarly, split A into  $A_0$  and  $A_1$

# How to make an atom? (2)

**①** Let  $\lambda$  be a measure on X. Split  $\lambda$  into  $\hat{\lambda}$  on  $\hat{X}$  such that

$$\lambda^*(A_0) = \lambda(A \cap C)(1 - \delta) + \lambda(A \cap C^c)$$

$$\lambda^*(A_1) = \lambda(A \cap C)\delta$$

- lacksquare In other words,  $\lambda^*(A_0 \cup A_1) = \lambda(A)$
- **3** Split transition kernel P into  $\hat{P}$ :

$$\hat{P}(x,\cdot) = P(x,\cdot)^* \quad x \in X_0 \setminus C_0$$

$$\hat{P}(x,\cdot) = (1-\delta)^{-1} [P(x,\cdot)^* - \delta \nu^*(\cdot)] \quad x \in C_0$$

$$\hat{P}(x,\cdot) = \nu^*(\cdot) \quad x \in C_1$$



# How to make an atom? (3)

- **1** A Markov process  $\hat{\Phi}_n$  is defined on  $\hat{X}$  with transition probability  $\hat{P}$ .
- $\circ$   $C_1$  becomes an atom.
- 0 Most result (irreducibility, aperiodicity, recurrence etc. ) still holds for  $\hat{\Phi}_n$

#### First simultaneous renewal time?

- $Y_1, Y_1, Y_2, Y_2, \cdots$  are i.i.d. with distribution  $\eta_{\alpha} \|_{\Phi_0 = \alpha}$
- Let T be the simultaneous renewal time

$$T = \inf_{n} \{ n = S_{k_1} = S'_{k_2} \text{ for some } k_1, k_2 \}$$

**o** From renewal theorem: There exist  $n_0$  and c such that

$$\mathbb{P}[n \text{ is a renewal time }] = \mathbb{P}[n = S_k \text{ for some } k] \geq c$$
 for all  $n > n_0$ .



#### Theorems

#### Exponential tail

If  $\mathbb{E}[\rho_1^{Y_0}], \mathbb{E}[\rho_1^{Y_0}], \mathbb{E}[\rho_1^{Y_1}] < \infty$  for some  $\rho_1 > 1$ , then there exists  $\rho_0 > 1$  such that  $\mathbb{E}[\rho_0^T] < \infty$ .

#### Power-law tail

If  $\mathbb{E}[Y_0^{\beta}], \mathbb{E}[(Y_0)^{\beta}], \mathbb{E}[Y_1^{\beta}] < \infty$  for some  $\beta > 0$ , then  $\mathbb{E}[T^{\beta}] < \infty$ .

(Note that finite exponential/power-law moment is equivalent to exponential/power-law tail.)

Proof on whiteboard.

Ref: Lectures on the Coupling Method by Torgny Lindvall



# Thank you