

1. Stochastic process

**Definition 0.1.** A stochastic process is a collection of random variables  $\{X_t\}_{t \in T}$ , where  $T$  is an index set.

2. Markov Property

**Definition 0.2.** A stochastic process  $\{X_t\}_{t \in T}$  has the Markov property if

$$P(X_{t_{n+1}} = x_{n+1} | X_{t_1} = x_1, \dots, X_{t_n} = x_n) = P(X_{t_{n+1}} = x_{n+1} | X_{t_n} = x_n) \quad (1)$$

3. Markov Chain

**Definition 0.3.** A Markov chain is a stochastic process with the Markov property.

4. Time-homogeneous Markov Chain

**Definition 0.4.** A Markov chain is time-homogeneous if the transition probabilities are independent of time.

5. Transition Probability

**Definition 0.5.** The transition probability  $P_{ij}$  is the probability of transitioning from state  $i$  to state  $j$ .

6. Transition Probability Matrix

**Definition 0.6.** The transition probability matrix  $P$  is a matrix where  $P_{ij}$  is the probability of transitioning from state  $i$  to state  $j$ .

7. n-step Transition Probability Matrix

**Definition 0.7.** The  $n$ -step transition probability matrix  $P^{(n)}$  is a matrix where  $P_{ij}^{(n)}$  is the probability of transitioning from state  $i$  to state  $j$  in  $n$  steps.

$$P_{ij}^{(n)} = P(X_{t_n} = j | X_{t_0} = i) \quad (2)$$

where  $t_n = t_0 + n$ .

8. Chapman-Kolmogorov Equation

**Theorem 0.1.** The Chapman-Kolmogorov equation is given by

$$P_{ij}^{(n+m)} = \sum_k P_{ik}^{(n)} P_{kj}^{(m)} \quad (3)$$

9. Hitting Probability

**Definition 0.8.** The hitting probability  $f_{ij}$  is the probability of reaching state  $j$  in the future starting from state  $i$ .

$$P_i(\text{hit } j) = f_{ij} = P(\text{reach state } j | \text{start at state } i) \quad (4)$$

$$\sum_{n=1}^{\infty} P(\text{first reach state } j \text{ in } n \text{ steps} | \text{start at state } i) \quad (5)$$

10. First Visit Probability

**Definition 0.9.** The first visit probability  $f_{ij}^{(n)}$  is the probability of reaching state  $j$  in exactly  $n$  steps starting from state  $i$ .

$$f_{ij}^{(n)} = P(\text{reach state } j \text{ in exactly } n \text{ steps} | \text{start at state } i) \quad (6)$$

11. Hitting Time

**Definition 0.10.** Given a stochastic path  $\{X_t\}_{t \in T}$ , the hitting time  $T_i^j$  is the first time the path reaches state  $j$  starting from state  $i$ .

We use  $T^j$  to denote the hitting time of state  $j$  starting from the current state.

12. Communicate

**Definition 0.11.** Two states  $i$  is said to Communicate with  $j$  and is written as  $i \rightarrow j$  if there exists a natural number  $n$  such that  $P_{ij}^{(n)} > 0$ .

13. Inter Communicate

**Definition 0.12.** Two states  $i$  and  $j$  are said to Inter Communicate and are written as  $i \leftrightarrow j$  if  $i \rightarrow j$  and  $j \rightarrow i$ .

**Example 0.1.**  $i \leftrightarrow i$  as  $P_{ii}^{(0)} = 1$ .

14. Irreducible Class of States

**Definition 0.13.** A class of states  $C = \{i_1, i_2, \dots, i_n\}$  is said to be Irreducible if for all  $i, j \in C$ ,  $i \leftrightarrow j$ .

15. Closed Class of States

**Definition 0.14.** A class of states  $C = \{i_1, i_2, \dots, i_n\}$  is said to be Closed if for all  $i \in C$ , there is no  $j \notin C$  such that  $i \rightarrow j$ .

16. Absorbing State

**Definition 0.15.** A state  $i$  is said to be Absorbing if  $P_{ii} = 1$ .

17. Recurrent (Persistent) State

**Definition 0.16.** A state  $i$  is said to be Recurrent or Persistent if the hitting probability  $f_{ii} = 1$ .

18. Transient State

**Definition 0.17.** A state  $i$  is said to be Transient if it is not Recurrent.

19. Recurrent Markov Chain

**Definition 0.18.** A Markov chain is said to be Recurrent if all states are Recurrent.

20. Transient Markov Chain

**Definition 0.19.** A Markov chain is said to be Transient if all states are Transient.

21. Identification of Recurrent States

**Theorem 0.2.** A state  $i$  is Recurrent if and only if the series

$$\sum_{n=1}^{\infty} P_{ii}^{(n)} = \infty \quad (7)$$

**Theorem 0.3.** If a state  $i$  is Transient, then

$$\lim_{n \rightarrow \infty} P_{ji}^{(n)} = 0, \forall j \quad (8)$$

Note, that the converse of this theorem is not true.

22. There is at least one Recurrent State for finite state space

**Theorem 0.4.** For a finite class of closed state space, there is at least one Recurrent State.

*Proof.* Let  $I$  be a finite class of closed state space. If all states are Transient, then

$$\lim_{n \rightarrow \infty} P_{ii}^{(n)} = 0, \forall i \in I \quad (9)$$

Given some  $i \in I$ , we have

$$P_{ii}^{(n)} = \sum_{j \in I} P_{ij} P_{ji}^{(n-1)} \quad (10)$$

For all  $j \in I$  such that  $P_{ij} > 0$ , we have

$$\lim_{n \rightarrow \infty} P_{ji}^{(n-1)} = 0 \quad (11)$$

if not, equation 10 will not hold.

Thus, for all  $i \in I$ , and all  $j \in I$  such that  $P_{ij} > 0$ , we have

$$\lim_{n \rightarrow \infty} P_{ii}^{(n)} = 0 \quad (12)$$

Then, for all  $n > 1$  we have

$$1 = \sum_{j \in I} P_{ij}^{(n)} \quad (13)$$

$$\lim_{n \rightarrow \infty} 1 = \lim_{n \rightarrow \infty} \sum_{j \in I} P_{ij}^{(n)} \quad (14)$$

$$1 = \sum_{j \in I} \lim_{n \rightarrow \infty} P_{ij}^{(n)} \quad (15)$$

$$1 = \sum_{j \in I} 0 \quad (16)$$

$$1 = 0 \quad (17)$$

This is a contradiction. Thus, there is at least one Recurrent State.  $\square$

23. Inter Communicate Classes have the same Recurrence

**Theorem 0.5.** *If two states  $i$  and  $j$  Inter Communicate, then  $i$  is Recurrent if and only if  $j$  is Recurrent.*

24. Number of visits to a state

**Definition 0.20.** *Given a stochastic path  $\{X_t\}_{t \in T}$ , the number of visits to state  $j$  is given by*

$$N_j = \sum_{n=1}^{\infty} 1_{\{X_n=j\}} \quad (18)$$

25. Number of visits to a state at infinity

**Theorem 0.6.**

$$P(N_i = \infty) = 1, \forall i \text{ Recurrent} \quad (19)$$

$$P(N_i = \infty) = 0, \forall i \text{ Transient} \quad (20)$$

26. Decomposition of State Space

**Theorem 0.7.** *The state space can be decomposed into a set of Irreducible Closed Recurrent classes and a set of Transient classes.*

*Proof.* Given state space  $I$ , as the InterCommunicate relation is an equivalence relation, we have a partition of  $I$  into Inter Communicate classes. Let the classes be  $I_\alpha, \alpha \in A$ .

As all states in an InnerCommunicate class have the same Recurrence, let

$$B = \{\alpha \in A | I_\alpha \text{ is Transient}\} \quad (21)$$

Let  $C = A - B$ . Then, for all  $\alpha \in C$ ,  $I_\alpha$  is Recurrent.

We then prove that  $I_\alpha$  is Closed for all  $\alpha \in C$ , which finished the proof of this theorem.

If  $I_\alpha$  is not closed, then there exists  $i \in I_\alpha$  and  $j \notin I_\alpha$ , and  $n \in \mathbb{N}$  such that  $P_{ij}^{(n)} > 0$ .

Also, as  $i \in I_\alpha$ ,  $i$  is Recurrent, and thus

$$f_{ii} = 1 \quad (22)$$

Also, in this case,  $j$  can not Communicate with  $i$ , as if it does,  $j$  will be in  $I_\alpha$ , which means

$$f_{ji} = 0 \quad (23)$$

Then we have

$$\begin{aligned}
1 &= f_{ii} \\
&= \sum_{k \in I} P_{ik}^{(n)} f_{ki} \\
&= P_{ij}^{(n)} * f_{ji} + \sum_{k \in I, k \neq j} P_{ik}^{(n)} f_{ki} \\
&= \sum_{k \in I, k \neq j} P_{ik}^{(n)} f_{ki} \\
&\leq \sum_{k \in I, k \neq j} P_{ik}^{(n)} \\
&= 1 - P_{ij}^{(n)} \\
&< 1
\end{aligned}$$

This is a contradiction. Thus,  $I_\alpha$  is closed.  $\square$

## 27. Mean Recurrence Time

**Definition 0.21.** *The mean recurrence time  $\mu_i$  is the expected number of steps to return to state  $i$  starting from state  $i$ .*

$$\mu_i = \begin{cases} \sum_{n=1}^{\infty} n f_{ii}^{(n)} & \text{if } i \text{ is Recurrent} \\ \infty & \text{if } i \text{ is Transient} \end{cases} \quad (24)$$

## 28. Generating Function of Mean Recurrence Time

**Definition 0.22.** *Define*

$$F_{ij}(s) = \sum_{n=1}^{\infty} f_{ij}^{(n)} s^n \quad (25)$$

*Then we have*

$$\mu_i = F'_{ii}(1) \quad (26)$$

## 29. Positive Recurrent and Null Recurrent

**Definition 0.23.** *A state  $i$  is said to be Positive Recurrent if*

$$\mu_i < \infty \quad (27)$$

*A state  $i$  is said to be Null Recurrent if*

$$\mu_i = \infty \quad (28)$$

## 30. Identification of Positive Recurrent States

**Theorem 0.8.** *A state  $i$  is Null Recurrent if and only if*

$$\lim_{n \rightarrow \infty} P_{ii}^{(n)} = 0 \quad (29)$$

31. Item in the same Irreducible Class have the same Recurrence

**Theorem 0.9.** *If two states  $i$  and  $j$  are in the same Irreducible class, then  $i$  is Positive Recurrent if and only if  $j$  is Positive Recurrent.*

*$i$  is Null Recurrent if and only if  $j$  is Null Recurrent.*

32. Finite State Class has at least one Positive Recurrent State

**Theorem 0.10.** *For a finite class of closed state space, there is at least one Positive Recurrent State.*

33. The Expected First Hit Time

**Definition 0.24.** *The expected first hit time  $\beta_{ij}$  is the expected number of steps to reach state  $j$  starting from state  $i$ .*

$$\beta_{ij} = \begin{cases} \sum_{n=1}^{\infty} n f_{ij}^{(n)} & \text{if } i \neq j \\ \mu_i & \text{if } i = j \end{cases} \quad (30)$$

34. Periodicity

**Definition 0.25.** *A state  $i$  is said to be periodic if the greatest common divisor of the set*

$$\{n \in \mathbb{N} | P_{ii}^{(n)} > 0\} \quad (31)$$

*is greater than 1.*

*If the greatest common divisor is 1, then the state is said to be aperiodic.*

35. State in Irreducible State Class Have Same Periodicity

**Theorem 0.11.** *Given two states  $i \leftrightarrow j$ , then  $i$  and  $j$  have the same period.*

36. Equilibrium (Stationary) Distribution of a Markov Chain

**Definition 0.26.** *A distribution  $\pi$  is said to be an equilibrium distribution if*

$$\sum_{i \in S} \pi_i = 1 \quad (32)$$

$$\pi = \pi P \quad (33)$$

37. Relationship Between Stationary State and Mean Recurrence Time

**Theorem 0.12.** *An Irreducible Markov Chain is Positive Recurrent if and only if there exists a unique stationary distribution  $\pi$ .*

*In this case, the stationary distribution is given by*

$$\pi_i = \frac{1}{\mu_i} \quad (34)$$

38. Ergodic Markov Chain Limiting Distribution

**Definition 0.27.** A Markov Chain is said to be Ergodic if it is Irreducible, Positive Recurrent, and Aperiodic.

**Theorem 0.13.** For an Ergodic Markov Chain, the limiting distribution exists and is given by

$$\lim_{n \rightarrow \infty} P_{ij}^{(n)} = \pi_j, \forall i, j \quad (35)$$

39. Passage Time to a Closed State Space

**Definition 0.28.** Given a closed state space  $A$ , and given a path  $\{X_t\}_{t \in T}$ , where  $X_0 = i$  the first passage time to  $A$  is given by

$$T_i^A = \inf\{t > 0 | X_t \in A\} \quad (36)$$

40. Absorption Probability to a Closed State Space

**Definition 0.29.** Given a closed state space  $A$ , and given a path  $\{X_t\}_{t \in T}$ , where  $X_0 = i$  the absorption probability to  $A$  is given by

$$\alpha_i^A = P(T_i^A < \infty) \quad (37)$$

41. Mean Passage Time to a Closed State Space

**Definition 0.30.** Given a closed state space  $A$ , and given a path  $\{X_t\}_{t \in T}$ , where  $X_0 = i$  the mean passage time to  $A$  is given by

$$\beta_i^A = E[T_i^A] \quad (38)$$