

Lecture 12 : Limit Theorems for Markov Chains

1 Limit Theorems

Let $N_j(t)$ denote the number of transitions into state $j \in E$ up to time t . That is,

$$N_j(t) = \sum_{k=1}^t 1\{X_k = j\}.$$

Let $S_0 = 0$, then we can define the n th arrival instants of state j as a stopping time

$$S_n(j) = \inf\{k > S_{n-1}(j) : X_k = j\}.$$

From strong Markov property it follows that X is a regenerative process with regenerative sequence $S(j) = \{S_n(j) : n \in \mathbb{N}\}$. We can define the inter-renewal duration, the number of time steps to return to the state j as

$$T_n(j) = S_n(j) - S_{n-1}(j).$$

If $X_0 = j$ and j is recurrent, then $S(j)$ is a renewal process with the *iid* inter-arrival distribution,

$$P_j\{T_1(j) = k\} = f_{jj}^{(k)}, \quad k \in \mathbb{N}.$$

Let $\mu_{jj} = \mathbb{E}_j T_1(j)$ be the mean inter-arrival time for the renewal process. Then,

$$\mu_{jj} = \begin{cases} \infty & j \text{ transient,} \\ \sum_{k \in \mathbb{N}} k f_{jj}^{(k)} & j \text{ recurrent.} \end{cases}$$

If $X_0 = i \neq j$, for some $i \leftrightarrow j$ and j recurrent, then $S(j)$ is a delayed renewal process with first inter-arrival distribution

$$P_i\{T_1(j) = k\} = f_{ij}^{(k)}, \quad k \in \mathbb{N}.$$

The associated counting process $N_j(t)$ has the inverse relationship with the renewal process $S(j)$. From the renewal theory, we have the following results.

Proposition 1.1 (basic renewal theorem). *If $i \leftrightarrow j$, then $P_i \left\{ \lim_{n \in \mathbb{N}} \frac{N_j(n)}{n} = \frac{1}{\mu_{jj}} \right\} = 1$.*

Proposition 1.2 (elementary renewal theorem). *If $i \leftrightarrow j$, then*

$$\lim_{n \in \mathbb{N}} \frac{\sum_{k=1}^n P_{ij}^{(k)}}{n} = \lim_{n \in \mathbb{N}} \frac{\mathbb{E}_i[N_j(n)]}{n} = \frac{1}{\mu_{jj}}.$$

Proposition 1.3 (Blackwell's theorem). *If j is aperiodic (i.e., $d(j) = 1$), then*

$$\lim_{n \in \mathbb{N}} p_{ij}^{(n)} = \lim_{n \in \mathbb{N}} \mathbb{E}_i[\# \text{ renewals at } n] = \frac{1}{\mu_{jj}}$$

If j is periodic with period d , then

$$\lim_{n \in \mathbb{N}} p_{ij}^{(nd)} = \lim_{n \in \mathbb{N}} \mathbb{E}_i[\# \text{ renewals at } nd] = \frac{d}{\mu_{jj}}.$$

2 Positive and Null recurrence

A recurrent state j is said to be **positive recurrent** if $\mu_{jj} < \infty$ and **null recurrent** if $\mu_{jj} = \infty$. Let

$$\pi_j \triangleq \lim_{n \in \mathbb{N}} p_{jj}^{(nd)},$$

where d is the period of state j . Then $\pi_j > 0$ if and only if j is positive recurrent and $\pi_j = 0$ if j is null-recurrent.

Proposition 2.1. *Positive recurrence and null recurrence are class properties.*

An state that is aperiodic and positive recurrent is called **ergodic**. For a homogeneous Markov chain on state space E with transition probability matrix E , a probability distribution $\{\pi_j : j \in E\}$ is said to be **stationary** if for all states $j \in E$

$$\pi_j = \sum_{k \in E} \pi_k P_{kj}.$$

More compactly, π is stationary if $\pi = \pi P$.

Observe that for a Markov chain starting with its stationary distribution, then the distribution remains invariant for all times. That is, if π is the stationary distribution, and the Markov chain has initial distribution $\nu(0) = \pi$ at time 0, then at any time $n \in \mathbb{N}$, the Markov chain has distribution $\nu(n) = \pi$. Moreover since X_n has discrete states in E , the finite collection $(X_n, X_{n+1}, \dots, X_{n+m})$ have the same joint distribution. Hence it is a stationary process, and for all $k, m \in \mathbb{N}, i \in E^k$

$$P\{X_1 = i_1, \dots, X_k = i_k\} = P\{X_{m+1} = i_1, \dots, X_{m+k} = i_k\}.$$

Theorem 2.2. *An irreducible, aperiodic Markov Chain with countable state space E is of one of the following types.*

i) *All the states are either transient or null recurrent. For all states $i, j \in E$,*

$$\lim_{n \in \mathbb{N}} P_{ij}^n = 0,$$

and there exists no stationary distribution.

ii) *All the states are positive recurrent, and hence the chain is ergodic. There exists a unique stationary distribution $\pi \in \Delta(E)$, defined for all $i, j \in E$*

$$\pi_j \triangleq \lim_{n \in \mathbb{N}} P_{ij}^n > 0.$$

Proof. Let $\{X_n : n \in \mathbb{N}\}$ be an irreducible, aperiodic Markov chain with countable state space E .

- i) Suppose that all states are either transient or null recurrent. Note that exactly one of these will hold since there is only one communicating class. This implies that $\mu_{jj} = \infty$ for each state $j \in E$, and it follows from Blackwell's theorem applied to renewals for Markov chains that for any states $i, j \in E$

$$\lim_{n \in \mathbb{N}} P_{ij}^{(n)} = \frac{1}{\mu_{jj}} = 0.$$

If there existed a stationary distribution $\pi \in \Delta(E)$ in this case. For any step size $n \in \mathbb{N}$ and states $i, j \in E$, we would then have

$$\pi_j = \sum_{i \in E} \pi_i P_{ij}^{(n)}, \quad P_{ij}^{(n)} \leq 1.$$

We can change limits and summation using dominated convergence theorem, to get for any $j \in E$

$$\pi_j = \sum_{i \in E} \pi_i \lim_{n \in \mathbb{N}} P_{ij}^{(n)} = 0.$$

This contradicts π being a stationary distribution, proving the first part of the theorem.

- ii) We assume that all states are positive recurrent. From the theorem hypothesis, elementary renewal theorem, and positive recurrence, we get

$$\pi_j = \lim_{n \in \mathbb{N}} P_{ij}^{(n)} = 1/\mu_{jj} > 0.$$

Further, for any finite set $A \subseteq E$, we have

$$\sum_{j \in A} P_{ij}^{(n)} \leq \sum_{j \in E} P_{ij}^{(n)} = 1.$$

Taking limit $n \in \mathbb{N}$ on both sides, we conclude that $\sum_{j \in A} \pi_j \leq 1$ for all A finite. Taking limit with respect to increasing sets $A \uparrow E$, we conclude,

$$\sum_{j \in E} \pi_j \leq 1.$$

Further, we can write for all $A \subseteq E$,

$$P_{ij}^{n+1} = \sum_{k \in E} P_{ik}^n P_{kj} \geq \sum_{k \in A} P_{ik}^n P_{kj}.$$

Applying limit $n \in \mathbb{N}$ on both sides, we get $\pi_j \geq \sum_{k \in A} \pi_k P_{kj}$ for all A finite. Hence, taking limits with respect to increasing sets $A \uparrow E$, we get for all state $j \in E$,

$$\pi_j \geq \sum_{k \in E} \pi_k P_{kj}.$$

Assuming that the inequality is strict for some state $j \in E$, we can sum the inequalities over all states $j \in E$. Since, summands are non-negative we can exchange summation orders to get

$$\sum_{j \in E} \pi_j > \sum_{j \in E} \sum_{k \in E} \pi_k P_{kj} = \sum_{k \in E} \pi_k \sum_{j \in E} P_{kj} = \sum_{k \in E} \pi_k.$$

This is a contradiction. Therefore, for any state $j \in E$

$$\pi_j = \sum_{k \in E} \pi_k P_{kj}.$$

Defining normalized $w_j = \frac{\pi_j}{\sum_{k \in I} \pi_k}$, we see that $\{w_j : j \in E\}$ is a stationary distribution and so at least one stationary distribution exists. If the initial distribution of this positive recurrent Markov chain is a stationary distribution $\{\lambda_j : j \in E\}$, then for any finite subset $A \subseteq E$, we get

$$\lambda_j = \Pr\{X_n = j\} = \sum_{i \in E} P_{ij}^n \lambda_i \geq \sum_{i \in A} P_{ij}^n \lambda_i.$$

As before, we take limit $n \in \mathbb{N}$, followed by limit of increasing subsets $A \uparrow E$, to obtain

$$\lambda_j \geq \sum_{i \in E} \pi_j \lambda_i = \pi_j.$$

To show $\lambda_j \leq \pi_j$, we use the fact that $P_{ij}^n \leq 1$. Let $A \subseteq E$ be a finite subset, then

$$\lambda_j = \sum_{i \in I} P_{ij}^n \lambda_i = \sum_{i \in A} P_{ij}^n \lambda_i + \sum_{i \notin A} P_{ij}^n \lambda_i \leq \sum_{i \in A} P_{ij}^n \lambda_i + \sum_{i \notin A} \lambda_i.$$

Using our standard approach of taking limit $n \in \mathbb{N}$, followed by $A \uparrow E$, we obtain

$$\lambda_j \leq \sum_{i \in E} \pi_j \lambda_i = \pi_j.$$

□

Corollary 2.3. *An irreducible, aperiodic Markov chain defined on a finite state space E will be positive recurrent.*

Proof. Suppose that the Markov chain is not positive recurrent, then

$$\lim_{n \in \mathbb{N}} P_{ij}^{(n)} = 0.$$

Interchanging limit and finite summation gives

$$0 = \sum_{j \in E} \lim_{n \in \mathbb{N}} P_{ij}^{(n)} = \lim_{n \in \mathbb{N}} \sum_{j \in E} P_{ij}^{(n)} = 1.$$

This is a contradiction. Hence the above mentioned chain is positive recurrent. □

Corollary 2.4. *For an irreducible and aperiodic Markov chain with stationary distribution π on countable state space E , we have*

$$\mathbb{E}_j[T_1(j)] \triangleq \mathbb{E}[T_1(j) | X_0 = j] = \frac{1}{\pi_j}, \quad j \in E.$$

Further, we can define the number of visits to state i during one renewal duration $S_1(j)$ as

$$N_i(S_1(j)) = \sum_{k=1}^{S_1(j)} 1\{X_k = i\}.$$

Proposition 2.5. *For an aperiodic and irreducible Markov chain X with stationary distribution π on countable state space E , the mean number of visits to state i in one return to state j is given*

$$\mathbb{E}_j N_i(S_1(j)) = \frac{\pi_i}{\pi_j}.$$

Proof. Let $X_0 = j \in E$, then from renewal reward theorem for renewal sequence $S(i)$ and definition of π ,

$$\lim_{n \in \mathbb{N}} P_j\{X_n = i\} = \lim_{n \in \mathbb{N}} \frac{\sum_{k=1}^n 1\{X_k = i\}}{n} = \frac{1}{\mathbb{E}_i[T_1(i)]} = \pi_i.$$

Result follows from rewriting of the above expression for renewal sequence $S(j)$ as

$$\lim_{n \in \mathbb{N}} P_j\{X_n = i\} = \lim_{n \in \mathbb{N}} \frac{\sum_{k=1}^n 1\{X_k = i\}}{n} = \frac{\mathbb{E}_j \sum_{k=1}^{S_1(j)} 1\{X_k = i\}}{\mathbb{E}_j S_1(j)} = \frac{\mathbb{E}_j N_i(S_1(j))}{\mathbb{E}_j T_1(j)} = \pi_j \mathbb{E}_j N_i(S_1(j)).$$

□

2.1 Ergodic theorem for Markov Chains

Proposition 2.6. *Let $\{X_n : n \in \mathbb{N}_0\}$ be an irreducible, aperiodic, and positive recurrent Markov chain on countable state space E with stationary distribution π . Let $f : E \rightarrow \mathbb{R}$, such that $\sum_{i \in E} |f(i)| \pi_i < \infty$, that is f is integrable over E with respect to π . Then for any initial distribution of X_0 ,*

$$\lim_{n \in \mathbb{N}} \frac{1}{n} \sum_{i=1}^n f(X_i) = \sum_{i \in E} \pi_i f(i) \text{ almost surely.}$$

Proof. Fix $X_0 = i \in E$. Let $S(i)$ be sequence of successive instants at which state i is visited, with $S_0(i) = 0$. For all $p \geq 0$, let $R_{p+1} = \sum_{n=S_p(i)+1}^{S_{p+1}(i)} f(X_n)$ be the net reward earned at the end of cycle $(p+1)$. Each cycle forms a renewal. By the strong Markov property, these cycles are independent. At each of these stopping times, Markov chain is in state $i \in E$. Since $\mathbb{E}_i S_1(i) = 1/\pi_i$, we get from renewal reward theorem

$$\lim_{n \in \mathbb{N}} \frac{\sum_{k=1}^n R_k}{n} = \pi_i \mathbb{E}_i \left[\sum_{n=1}^{S_1(i)} f(X_n) \right] = \pi_i \mathbb{E}_i \sum_{n=1}^{S_1(i)} \sum_{j \in E} f(j) 1\{X_n = j\}.$$

Using dominated convergence theorem, and substituting the mean number of visits to state j during successive return to state j , we can write

$$\lim_{n \in \mathbb{N}} \frac{\sum_{k=1}^n R_k}{n} = \pi_i \mathbb{E}_i \sum_{j \in E} f(j) \sum_{n=1}^{S_1(i)} 1\{X_n = j\} = \pi_i \sum_{j \in E} f(j) \mathbb{E}_i N_j(S_1(i)) = \sum_{j \in E} \pi_j f(j).$$

□