# MATH 285A- Lecture Notes #8

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## 1 Warning

Invariant Measure :

$$\lambda = (\lambda_i)_{i \in S}$$
$$\lambda \ge 0, \quad \lambda \neq 0$$
$$\lambda' P = \lambda'.$$

Invariant Distribution: (Invariant probability distribution, stationary distribution)

$$\pi = (\pi_i)_{i \in S}$$
$$\pi_i \geq 0, \quad \forall i, \quad \sum_{i \in S} \pi_i = 1$$
$$\pi' P = \pi'$$

## 2 Hw Problem-last question (hint)

Show that the Markov chain in Exercise 1.3.4 is transient and deduce that  $P_i(X_n \to \infty \quad as \quad n \to \infty) = 1$  given  $X_0 = i$  for any  $i \ge 1$ .

We want to prove that

$$P_i(X_n \to \infty \quad as \quad n \to \infty) = 1, \quad \forall i \in S = \{0, 1, 2, ...\}$$
$$i.e \quad X_n \to \infty \quad as \quad n \to \infty \quad P_i - a.s, \quad \forall i$$
(1)

Fix i and focus on (1).

i.e., we want to prove that  $X_n(w) \to \infty$  as  $n \to \infty$  for almost every w under  $P_i(\forall i)$ .

 $+ - + - + - + - + - + - + - + - - time line 0 1 2 \dots k$ 

Want to show that for each  $k \ge 1$ ,  $X_n(w) > k$  for all n sufficiently large for almost every w under  $P_i$ , i.e.,  $X_n$  should visit states in [0,k] only finitely many times  $P_i$ -a.s.

Want to prove  $P_i(X_n \text{ is eventually not in } [0,k]) = 1$  for each  $k \ge 1, (\forall i)$ 

## **3** Strong Markov Property

X Markov process  $(\lambda, P)$ .

#### 3.1 Stopping Times (Optional Times)

A stopping time (with respect to X) is a function  $T : \Omega \to \{0, 1, 2...\} \cup \{\infty\}$  such that for each m = 0, 1, 2...

 ${T = m} = {w \in \Omega : T(w) = m}$  depends only on  $X_0, X_1, ..., X_m$ .

#### **Example:**

Fix state j:  $T^j = inf\{n \ge 1 : X_n = j\}$  is a stopping time.

 $\frac{\text{Check}}{\text{Fix } m \in \{0, 1, 2, ...\}}$ 

$$\{T^{j} = m\} = \emptyset, \qquad m = 0. \\ \{X_{1} \neq j, X_{2} \neq j, ..., X_{m-1} \neq j, X_{m} = j\} \quad m \ge 1.$$

The following is not in general a stopping time:  $T_i = sup\{n \ge 1 : X_n = i\} = last$  time visit i. To determine the event  $\{T_i = m\}$ , you need to know that happens to X after m, and so this is not a stopping time in general.

### 3.2 Strong Markov Property

Let *T* be a stopping time relative to the Markov chain *X*. Conditioned on  $T < \infty$  and  $X_T = i$ ,  $(X_{T+n})_{n=o}^{\infty}$  is a Markov chain with parameters  $(\delta_i, P)$ .  $\delta_i$  = point mass at i, i.e.,

$$P(X_{T+n} = j_n, X_{T+n-1} = j_{n-1}, \dots, X_T = j_0 \mid X_0 = i_0, X_1 = i_1, \dots, X_T = i, T < \infty).$$
(2)

some event in the future beyond T

event in the past up to  $T < \infty$ .

$$= P_i(X_n = j_n, X_{n-1} = j_{n-1}, \dots, X_0 = j_0), \quad \forall \ n \ge 0, j_0, j_1, \dots, j_n, \quad i_0, i_1, \dots i_n.$$

Proof:

$$\{T < \infty\} = \bigcup_{m=0}^{\infty} \{T = m\}$$

Want to prove (2). It is equivalent to :

$$P(X_{T+n} = j_n, X_{T+n-1} = j_{n-1}, \dots, X_T = j_0, X_0 = i_0, X_1 = i_1, \dots, X_T = i, T < \infty)$$
  
=  $P_i(X_n = j_n, X_{n-1} = j_{n-1}, \dots, X_0 = j_0) * P(X_0 = i_0, X_1 = i_1, \dots, X_T = i, T < \infty).$ 

Enough to show for each m = 0, 1, 2, ...

$$P(X_{T+n} = j_n, X_{T+n-1} = j_{n-1}, \dots, X_T = j_0; X_0 = i_0, X_1 = i_1, \dots, X_T = i, T = m)$$
  
=  $P_i(X_n = j_n, X_{n-1} = j_{n-1}, \dots, X_0 = j_0) * P(X_0 = i_0, X_1 = i_1, \dots, X_T = i, T = m).$  (3)

In time subscripts, can replace *T* by *m* in the above on  $\{T = m\}$  and then (3) follows by regular Markov property.  $\Box$ 

## 3.3 Stationary Distribution

Assume M.C. is irreducible and positive recurrent (not necessarily aperiodic). Let

$$T_k = \inf\{n \ge 1 : X_n = k\}$$

 $T_k$ : time to return to k.

$$\gamma_i^k = E_k \bigg[ \sum_{n=0}^{T_k-1} \mathbf{1}_{\{X_n=i\}} \bigg], \quad k \in \mathbb{S}, i \in \mathbb{S}$$

 $\gamma_i^k$ : amount of time you spend in i before come back to k. By Theorem 1.7.5 we have the following:  $\gamma_k^k=1$  and for

$$\gamma^k = \begin{pmatrix} \gamma_0^k \\ \gamma_1^k \\ \vdots \\ \vdots \\ \vdots \end{pmatrix},$$

$$\gamma^k P = \gamma^k, \quad (invariant).$$

Since *X* is positive recurrent,

$$\sum_{i \in \mathbb{S}} \gamma_i^k = \sum_{i \in \mathbb{S}} E_k \left[ \sum_{n=0}^{T_k - 1} \mathbb{1}_{\{X_n = i\}} \right]$$

$$= E_k \left[ \sum_{i \in \mathbb{S}} \sum_{n=0}^{T_k - 1} \mathbb{1}_{\{X_n = i\}} \right]$$

$$= E_k \left[ \sum_{n=0}^{T_k - 1} \sum_{i \in \mathbb{S}} \mathbb{1}_{\{X_n = i\}} \right]$$

$$= E_k \left[ \sum_{n=0}^{T_k - 1} \mathbb{1} \right]$$

$$= E_k \left[ T_k \right] = m_k < \infty \quad (because positive recurrent)$$
(5)

So,

$$\pi_i = \frac{\gamma_i^k}{m_k}, \quad i \in \mathbb{S}.$$

is a stationary distribution. (normalized for elements to sum to 1.)

Irreducibility gives that  $\pi$  is unique (needs an argument - see text). Then we must have  $\pi_k = \frac{1}{m_k}$  for each  $k \in S$ , is the unique stationary distribution.

#### 3.4 ERGODIC Thm:

Assume M.C. is positive recurrent and irreducible.

Fix starting state *i*.

Fix state  $k \in \mathbb{S}$ ,

$$T_0^k = inf\{n \ge 0 : X_n = k\}$$

first time, X is in k.

 $T_1^k$ : amount of time between the first and 2nd visits to k.  $T_2^k$ : amount of time between the 2nd and 3rd visits to k..... Let

$$W_n^k = \sum_{l=0}^n T_l^k$$

waiting time till the (n + 1)th visit to k. Let

$$W_{-1}^{k} = 0$$

By the strong Markov property {  $T_l^k$  } $_{l=1}^{\infty}$  are *i.i.d* with finite mean  $m_k = E_i [T_1^k]$ .

#### S.L.L.N (Strong Law of Large Nos)

 $P_i - a.s.$ 

$$\frac{1}{n}\sum_{l=1}^{n} T_{l}^{k} \to m_{k} \quad as \quad n \to \infty.$$

So,

$$\frac{W_n^k}{n} \to m_k \quad as \quad n \to \infty, \quad P_i - a.s$$

What was the time of the last visit to k before n?

$$V_k(n) = \sum_{l=0}^{n-1} 1_{\{X_l=k\}} \quad \to \infty \quad as \ n \to \infty \quad (by \ recurrence)$$

 $V_k(n)$  : number of visits to k up to time n-1.

$$W_{V_k(n)-1}^k \le n - 1 < W_{V_k(n)}^k$$

so

$$W_{V_k(n)-1}^k < n \le W_{V_k(n)}^k$$

Divide by  $V_k(n)$ ,

$$\frac{W_{V_k(n)-1}^k}{V_k(n)} < \frac{n}{V_k(n)} \le \frac{W_{V_k(n)}^k}{V_k(n)}$$

Both the extreme left and extreme right expressions tend to  $m_k P_i$ -a.s., as  $n \to \infty$ , because  $V_k(n)$  tends to infinity  $P_i$ -a.s. as  $n \to \infty$ , and  $W_{\ell}^k/\ell$  tends to  $m_k P_i$ -a.s. as  $\ell$  tends to infinity, hence the composition tends to  $m_k$ ,  $P_i$ -a.s. as  $n \to \infty$ .

Thus,

$$\frac{n}{V_k(n)} \to m_k \quad as \quad n \to \infty \quad P_i - a.s.$$

$$rac{V_k(n)}{n} 
ightarrow rac{1}{m_k} \quad as \quad n 
ightarrow \infty \quad P_i - a.s.$$