

Lecture 11:

Proof of the Convergence Theorem I, Uniqueness

Assumptions:

I: Irreducible

A: Aperiodic

R: Recurrent

S: existence of a stationary distribution $\vec{\pi}$

Lemma 11.1. If S holds, then all states y that have $\vec{\pi}_y > 0$ are recurrent.

Proof. From Lemma 5 & Lemma 6 in Lecture 7, we know two representations of $E_x N(y)$:

$$E_x N(y) = \sum_{n=1}^{\infty} [P^n]_{xy} = \begin{cases} 0, & P_{xy} = 0; \\ \frac{P_{xy}}{1 - P_{yy}}, & P_{xy} > 0. \end{cases}$$

On the one hand, since $\pi_y > 0$, one has

$$\begin{aligned} \sum_{x \in \mathcal{X}} \pi_x E_x N(y) &= \sum_{x \in \mathcal{X}} \sum_{n=1}^{\infty} \pi_x [P^n]_{xy} = \sum_{n=1}^{\infty} \sum_{x \in \mathcal{X}} \pi_x [P^n]_{xy} \\ &= \sum_{n=1}^{\infty} [\vec{\pi} P^n]_y = \sum_{n=1}^{\infty} \vec{\pi}_y = \infty \cdot \vec{\pi}_y = \infty. \end{aligned}$$

On the other hand, we have

$$\sum_{x \in \mathcal{X}} \pi_x E_x N(y) \leq \sum_{x \in \mathcal{X}} \pi_x \cdot \frac{1}{1 - P_{yy}} = \frac{1}{1 - P_{yy}}.$$

Therefore, $\frac{1}{1 - P_{yy}} = \infty$ and thus $P_{yy} = 1$ ■.

Remark 11.1. If I & S hold, then R also holds.

Proof. Since $\sum_{x \in \mathcal{X}} \bar{\pi}_x = 1$, there exists $y \in \mathcal{X}$ with $\bar{\pi}_y > 0$.

why?

Lemma 11.1 says y is recurrent. Thus, R holds.

Theorem 9.3. If I & S hold, then $\bar{\pi}_y = \frac{1}{E_y \tau_y}$, $\forall y \in \mathcal{X}$.

Proof. Suppose $X_0 \sim \bar{\pi}$. From Theorem 9.2, it follows that

$$\frac{N_n(y)}{n} \xrightarrow{\text{a.s.}} \frac{1}{E_y \tau_y}, \quad \forall y \in \mathcal{X}.$$

Taking expected values at both sides, one has

$$E_{\bar{\pi}} E_x \frac{N_n(y)}{n} \xrightarrow{n \rightarrow \infty} \frac{1}{E_y \tau_y}, \quad \forall y \in \mathcal{X}.$$

Notice that $N_n(y) = \sum_{i=1}^n \mathbb{1}_{\{X_i=y\}}$, and

$$E_x N_n(y) = \sum_{i=1}^n E_x \mathbb{1}_{\{X_i=y\}} = \sum_{i=1}^n P_x(X_i=y) = \sum_{i=1}^n [P^i]_{xy}.$$

$$\text{Thus, } E_{\bar{\pi}} E_x N_n(y) = \sum_{x \in \mathcal{X}} E_x N_n(y) \bar{\pi}_x$$

$$= \sum_{x \in \mathcal{X}} \sum_{i=1}^n [P^i]_{xy} \cdot \bar{\pi}_x$$

$$= \sum_{i=1}^n [\bar{\pi} \cdot P^i]_y = \sum_{i=1}^n \bar{\pi}_y = n \bar{\pi}_y.$$

Therefore, $\vec{\pi}_y = \frac{1}{E_y \tau_y}$, $\forall y \in \mathcal{X}$. \blacksquare

Recall: Theorem 8.1. If I & R hold and $\vec{\mu}_y^x := \sum_{n=0}^{\infty} P_x(X_n=y, \tau_x > n)$, then $\vec{\mu}^x$ is a stationary measure with $0 < \vec{\mu}_y^x < \infty$, $\forall y \in \mathcal{X}$.

Proposition 11.1. If I , R & S hold and $\vec{\mu}_y^x := \sum_{n=0}^{\infty} P_x(X_n=y, \tau_x > n)$, then $\vec{\mu}_y^x = \frac{\pi_y}{\pi_x}$, $\forall x, y \in \mathcal{X}$.

Proof. For any fixed $x \in \mathcal{X}$, Theorem 8.1 implies that $\vec{\pi}_x \cdot \vec{\mu}^x$ is a stationary measure. Notice that

$$\begin{aligned} \sum_{y \in \mathcal{X}} \vec{\mu}_y^x &= \sum_{y \in \mathcal{X}} \sum_{n=0}^{\infty} P_x(X_n=y, \tau_x > n) \\ &= \sum_{n=0}^{\infty} \sum_{y \in \mathcal{X}} P_x(X_n=y, \tau_x > n) \\ &= \sum_{n=0}^{\infty} P_x(\tau_x > n) \\ &= E_x \tau_x \end{aligned}$$

why?

By Thm 9.3

$$= \frac{1}{\vec{\pi}_x}$$

Thus, $\sum_{y \in \mathcal{X}} [\bar{\pi}_x \cdot \bar{\mu}^x]_y = \bar{\pi}_x \cdot \sum_{y \in \mathcal{X}} \bar{\mu}_y^x = 1$.

Therefore, $\bar{\pi}_x \cdot \bar{\mu}^x$ is also a stationary distribution.

Then Corollary 9.1 implies $\bar{\pi}_x \cdot \bar{\mu}^x = \bar{\pi}$.

That is, $\forall y \in \mathcal{X}$, $\bar{\pi}_y = [\bar{\pi}_x \cdot \bar{\mu}^x]_y = \bar{\pi}_x \cdot \bar{\mu}_y^x$. ■

Recall Theorem 9.4. If I & S hold, and $f: \mathcal{X} \rightarrow \mathbb{R}$ has

$\sum_{x \in \mathcal{X}} |f(x)| \bar{\pi}_x < \infty$, then

$$\frac{1}{n} \sum_{k=0}^{n-1} f(X_k) \xrightarrow{\text{a.s.}} \sum_{x \in \mathcal{X}} f(x) \bar{\pi}_x = \mathbb{E}_{X \sim \bar{\pi}} [f(X)].$$

Sketch of the Proof. Suppose the chain starts at

x . Let $\tau_x^0 := 0$, and $\tau_x^k := \min\{n > \tau_x^{k-1} : X_n = x\}$ be the

time of the k -th return to x . Define

$$Y_k := \sum_{m=\tau_x^{k-1}+1}^{\tau_x^k} f(X_m), \quad \forall k=1, 2, \dots$$

By the Strong Markov Property, Y_1, Y_2, Y_3, \dots are

i.i.d. Then the Strong Law of Large Numbers tells

$$\frac{1}{L} \sum_{k=1}^L Y_k \xrightarrow{\text{a.s.}} \mathbb{E}_x Y_1. \quad (*)$$

On the other hand,

$$\mathbb{E}_x Y_1 = \mathbb{E}_x \left[\sum_{m=1}^{\tau_x^1} f(X_m) \right]$$

$$= \mathbb{E}_x \left[\sum_{m=1}^{\infty} f(X_m) \cdot \mathbb{1}_{\{\tau_x^1 \geq m\}} \right]$$

$$= \sum_{m=1}^{\infty} \mathbb{E}_x [f(X_m) \cdot \mathbb{1}_{\{\tau_x^1 \geq m\}}]$$

$$= \sum_{m=1}^{\infty} \sum_{y \in \mathcal{X}} f(y) \mathbb{P}_x(X_m = y, \tau_x^1 \geq m)$$

$$= \sum_{y \in \mathcal{X}} f(y) \sum_{m=1}^{\infty} \mathbb{P}_x(X_m = y, \tau_x^1 \geq m)$$

$$= \sum_{y \in \mathcal{X}} f(y) \sum_{m=0}^{\infty} \mathbb{P}_x(X_m = y, \tau_x^1 > m)$$

$$= \sum_{y \in \mathcal{X}} f(y) \bar{\mu}_y^x$$

$$= \sum_{y \in \mathcal{X}} f(y) \cdot \frac{\bar{\pi}_y}{\bar{\pi}_x}.$$

why?

Notice that $\sum_{k=1}^{N_n(x)} Y_k \leq \sum_{m=1}^n f(X_m) < \sum_{k=1}^{N_n(x)+1} Y_k$.

Taking $L = N_n(x) = \max\{k: \tau_x^k \leq n\}$, one has

$$\frac{N_n(x)}{n} \cdot \frac{1}{L} \sum_{k=1}^L Y_k \leq \frac{1}{n} \sum_{m=1}^n f(X_m) < \frac{N_n(x)+1}{n} \cdot \frac{1}{L+1} \sum_{k=1}^{L+1} Y_k.$$

From (*) and Corollary 9.2, we have

$$\frac{1}{n} \sum_{m=1}^n f(X_m) \xrightarrow{\text{a.s.}} \bar{\pi}_x \cdot \sum_{y \in \mathcal{X}} f(y) \cdot \frac{\bar{\pi}_y}{\bar{\pi}_x} = \sum_{y \in \mathcal{X}} f(y) \cdot \bar{\pi}_y.$$

Theorem 11.1 If I holds, and \vec{v} satisfies $\begin{cases} \vec{v} P = \vec{v} \\ \vec{v}_x = 1 \end{cases}$ for some x ,

then $\vec{v} = \vec{\mu}^x$. Here $\vec{\mu}^x$ is defined in Theorem 8.1 such

that $\mu_y^x = \sum_{n=0}^{\infty} P_x(X_n = y, \tau_x > n) \quad \forall y \in \mathcal{X}$.

Proof. For any $y \neq x$, $v_y = \sum_{y_1 \in \mathcal{X}} v_{y_1} P_{y_1, y} = P_{xy} + \sum_{y_1 \neq x} v_{y_1} P_{y_1, y}$.

Using the above equation recursively, one has

$$\begin{aligned} v_y &= P_{xy} + \sum_{y_1 \neq x} (P_{xy_1} + \sum_{y_2 \neq x} v_{y_2} P_{y_2, y_1}) P_{y_1, y} \\ &= P_{xy} + \sum_{y_1 \neq x} P_{xy_1} P_{y_1, y} + \sum_{y_1, y_2 \neq x} (P_{xy_2} + \sum_{y_3 \neq x} v_{y_3} P_{y_3, y_2}) P_{y_2, y_1} P_{y_1, y} \\ &= \dots \\ &\geq P_{xy} + \sum_{y_1 \neq x} P_{xy_1} P_{y_1, y} + \sum_{y_1, y_2 \neq x} P_{xy_2} P_{y_2, y_1} P_{y_1, y} + \dots \\ &= P_x(X_1 = y, \tau_x > 1) + P_x(X_2 = y, \tau_x > 2) + P_x(X_3 = y, \tau_x > 3) + \dots \\ &= \mu_y^x, \quad \forall y \neq x. \end{aligned}$$

Also, $\vec{v}_x = 1 = \vec{\mu}_x^x$. Thus, $\vec{v} \geq \vec{\mu}^x$.

Define $\vec{w} = \vec{v} - \vec{\mu}^x$.

Then, $\vec{w} \geq 0$ and $\vec{w} P = \vec{v} P - \vec{\mu}^x P = \vec{v} - \vec{\mu}_x = \vec{w}$.

For any y , since P is irreducible, $\exists M_y$, s.t. $[P^{M_y}]_{yx} > 0$.

$$\begin{aligned} \text{Thus, } 0 = \vec{w}_x &= [\vec{w} \cdot P^{M_y}]_x = \sum_{z \in X} \vec{w}_z [P^{M_y}]_{zx} \geq \vec{w}_y \cdot [P^{M_y}]_{yx} \\ &\geq 0 \end{aligned}$$

This implies, $\vec{w}_y \cdot [P^{M_y}]_{yx} = 0$.

And thus, $\vec{w}_y = 0$, $\forall y \in X$.

Therefore, $\vec{v} = \vec{\mu}_x$. \square

Theorem 11.2. If I & R hold, all the stationary measures are unique up to a constant.

Proof. Suppose $\vec{\mu}$ is a stationary measure. There exists $x \in X$, s.t. $\vec{\mu}_x > 0$. Let $\vec{v} := \frac{\vec{\mu}}{\vec{\mu}_x}$. Then $\vec{v} = \vec{v} P$, and $\vec{v}_x = 1$. Theorem 11.1 tells $\vec{v} = \vec{\mu}^x$.

Thus, $\bar{\mu} = \bar{\mu}_x \cdot \bar{\mu}^x$. This means, any stationary measure equals to an element from $\{\bar{\mu}^x: x \in \mathcal{X}\}$, up to some positive constant. For any $y \in \mathcal{X}$, Theorem 8.1 says $0 < \bar{\mu}_x^y < \infty$. Thus, $\bar{\mu}^y = \bar{\mu}_x^y \cdot \bar{\mu}^x$. Therefore, any stationary measure equals to $\bar{\mu}^x$, up to some positive constant. □

This is the end of this lecture !