

Föreläsning 20/11-13

6.4 Stationary distribution & the limit theorem

π (row matrix) is a stationary distribution for a Markov chain $(X_n; n \geq 0)$ with transition matrix P if π is a discrete probability distribution (that is, $\pi_i \geq 0$ and $\sum \pi_i = 1$) and $\pi P = \pi$.

Thm

If $\mu^{(0)} = \pi$
 If $\mu^{(n)} = \pi$ (remember $\mu_i^{(n)} = P(X_n = i)$) then $\mu^{(n)} = \pi$ for all n .
 $\mu^{(n+m)} = \pi$ for all n .

Proof: $\mu^{(n)} = \underbrace{\mu^{(0)}}_{\pi} P^n = \underbrace{(\pi P)}_{\pi} P^{n-1} = \dots = \pi$ \square

THEOREM

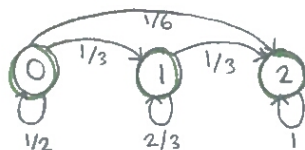
An irreducible chain has a stationary distribution π iff all states are non-null persistent. Then $\pi_i = 1/\mu_i$.
 ($\mu_i = E(T_i | X_0 = i)$, $T_i = \min(n \geq 1 : X_n = i)$)

Computer problem

Markov chain $(X_n; n \geq 0)$ with possible values $\{0, 1, 2\}$.
 $\mu^{(0)} = (1 \ 0 \ 0)$

$$P = \begin{bmatrix} 1/2 & 1/3 & 1/6 \\ 0 & 2/3 & 1/3 \\ 0 & 0 & 1 \end{bmatrix}$$

Find $E(T)$ where $T = \min(n \geq 1 : X_n = 2)$.



①, ② transient
 ② persistent

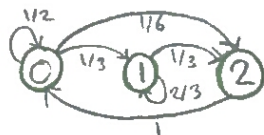
Two analytical solutions of ditto:

$$\begin{aligned} 1) E(T) &= E(\text{of time until we leave } \textcircled{0}) + \frac{2}{3} E(\text{of time until we leave } \textcircled{1}) = \\ &= \underbrace{\sum_{n=1}^{\infty} \frac{1}{2}^{n-1} \frac{1}{2} n}_{=2} + \frac{2}{3} \underbrace{\sum_{n=1}^{\infty} \frac{2}{3}^{n-1} \frac{1}{3} n}_{=3} = 2 + 3 \cdot \frac{2}{3} = 4 \end{aligned}$$

$$\left(\sum_{n=1}^{\infty} n x^{n-1} = \frac{d}{dx} \sum_{n=1}^{\infty} x^n = \frac{d}{dx} \frac{x}{1-x} \right)$$

2) Consider

$$P = \begin{bmatrix} 1/2 & 1/3 & 1/6 \\ 0 & 2/3 & 1/3 \\ 1 & 0 & 0 \end{bmatrix}$$



$$E(T) = \mu_2 - 1 = \frac{1}{\pi_2} - 1 = \dots$$

↑
 the time it takes to come to 2 from 2.

→ forts.

$$(\pi_0 \ \pi_1 \ \pi_2) \begin{pmatrix} 1/2 & 1/3 & 1/6 \\ 0 & 2/3 & 1/3 \\ 1 & 0 & 0 \end{pmatrix} = (\pi_0 \ \pi_1 \ \pi_2)$$

$$\Leftrightarrow \begin{cases} \frac{1}{2} \pi_0 + \pi_2 = \pi_0 & \pi_2 = 1/2 \pi_0 \\ \frac{1}{3} \pi_0 + \frac{2}{3} \pi_1 = \pi_1 & \pi_1 = \pi_0 \\ \frac{1}{6} \pi_0 + \frac{1}{3} \pi_1 = \pi_2 & \pi_2 = 1/2 \pi_0 \\ \pi_0 + \pi_1 + \pi_2 = 1 & \pi_0 + \pi_0 + 1/2 \pi_0 = 5/2 \pi_0 = 1 \end{cases}$$

$$\Leftrightarrow \pi_0 = 2/5, \ \pi_1 = 2/5, \ \pi_2 = 1/5$$

THEOREM

For any state j of a Markov chain with period $d(j)=1$.
 $(d(j) = \text{GCD}\{n \geq 1: P_{jj}(n) > 0\})$ We have $P_{jj}(n) \rightarrow 1/\mu_j$ as $n \rightarrow \infty$
 and $P_{ij}(n) \rightarrow \frac{f_{ij}}{\mu_j}$ as $n \rightarrow \infty$ $i \neq j$

$$(f_{ij} = P(\sum_{n=1}^{\infty} \mathbb{1}_{\{X_n=j\}} \mid X_0=i))$$

6.3 Classification of chains

Thm

If $i \leftrightarrow j$ (=intercommunicates) then

- i and j have same period.
- i is transient iff j is transient
- i is null persistent iff j is null persistent.

Proof: There exists $m, n > 0$ such that $\alpha = P_{ij}(m) P_{ji}(n) > 0$
 Therefore $P_{ij}(m+r+n) \geq P_{ij}(m) P_{jj}(r) P_{ji}(n) = \alpha P_{jj}(r)$
 $P_{jj}(m'+r+n') \geq P_{ji}(m') P_{ij}(r) P_{jj}(n') = \alpha' P_{ij}(r)$

$$\sum_{r=1}^{\infty} P_{ij}(r) = \infty \iff \sum_{r=1}^{\infty} P_{jj}(r) = \infty, \quad i \text{ transient iff } \sum_{n=1}^{\infty} P_{ij}(n) < \infty$$



$$i \text{ persistent iff } \sum_{n=1}^{\infty} P_{ij}(n) = \infty$$

$$i \text{ null pers. iff pers with } \lim_{n \rightarrow \infty} P_{ij}(n) = 0$$

Definition

A set C of states is called

- closed if $P_{ij} = 0$ for all $i \in C, j \notin C$
- irreducible if $i \leftrightarrow j \ \forall ij \in C$.

Thm (decomposition)

The state space S (the set of possible values of the Markov chain) can be decomposed as $S = T \cup C_1 \cup C_2 \dots$ where T are the transient states and C_1, C_2, \dots are closed irreducible sets of persistent states.

↑ transient means $P(\sum_n = i \text{ some } n \geq 1 \mid \sum_0 = i) = f_i < 1$
 $P(\text{that you eventually "escape" from a transient state } i \text{ and never come back.})$

example

$$P = \begin{pmatrix} 1/2 & 1/2 & & & & & \\ 1/4 & 3/4 & & & & & \\ 1/4 & 1/4 & 1/4 & 1/4 & & & \\ 1/4 & 0 & 1/4 & 1/4 & 0 & 1/4 & \\ & & & & 1/2 & 1/2 & \\ & & & & 1/2 & 1/2 & \end{pmatrix} \quad \begin{array}{l} S = \{0, 1, 2, 3, 4, 5\} \\ T = \{2, 3\} \\ C_1 = \{0, 1\} \\ C_2 = \{4, 5\} \end{array}$$

Thm

If S (state space of possible values) is finite then at least one state is persistent and all persistent states are non-null.