

IEOR 4106 lec 7

- 1) More on irreducible
Positive recurrent
Markov chains
- 2) Why π is called a
Stationary distribution

Recall that if a MC (X_n) is irreducible, then if all states are positive Recurrent

$$\left(\begin{array}{l} P(\tau_{ij} < \infty) = 1 \\ \text{and } E(\tau_{ij}) < \infty \end{array} \right)$$

then a limiting probability dist.

$$\pi = (\pi_j)_{j \in \mathcal{A}} \text{ exists}$$

$$\text{and } \left[\pi_j = \frac{1}{E(\tau_{jj})} \right]_{j \in \mathcal{A}}$$

$$\textcircled{1} \pi_j \stackrel{\text{def}}{=} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{m=1}^n \mathbb{I}\{X_m = j \mid X_0 = i\}$$

wpl

= long-run proportion of time

↑
 (constant that doesn't depend on the initial state $X_0 = i \in \mathcal{S}$; the chain spends (visits) in state j)

and $\pi_j > 0, j \in \mathcal{S},$

$$\sum_{j \in \mathcal{S}} \pi_j = 1 \quad \text{a probability distribution}$$

Rates for understanding Π

Π_j = long-run prop. of time the chain enters state j

= long-run prop. of time the chain leaves state j

= rate at which the chain enters state j

= rate at which the chain leaves state j

from
Basic

Principles;
not
requiring
"Markov"

rate out of j

π_j

=

rate into j

$$\sum_{i \in S} \pi_i P_{ij}$$

(Markov Property)

$j \in S$

Matrix form

$$\pi = \pi P$$

For MCS

(π is a row vector)

$$\pi_j = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{m=1}^n P_{ij}^{(m)}$$

$$\left(E(\mathbb{I}\{X_m=j | X_0=i\}) = P_{ij}^{(m)} \right)$$

$$\begin{pmatrix} \pi \\ \pi \\ \vdots \\ \pi \end{pmatrix} = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{m=1}^n P^m \quad (2)$$

↑ each row is $\pi = (\pi_j)$

Multiply both sides of (2) by P

$$\begin{aligned} \left(\begin{array}{c} \pi \\ \vdots \\ \pi \end{array} \right) P &= \left(\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{m=1}^n P^{m+1} \right) \\ &= \lim_{n \rightarrow \infty} \left[\frac{1}{n} \sum_{m=1}^n P^m \right] + \frac{(P^{n+1} - P)}{n} \end{aligned}$$

$\pi =$

$\rightarrow \pi P = \pi$

for an irreducible $M \subset (X_n)$

it is positive rec.

if and only if there exists

a prob. soln π

to $\pi = \pi P$ ($\pi_j = \sum_{i \in A} \pi_i P_{ij}$, $j \in A$)
in which case

π is the limiting dist.,

$$\left(\pi_j = \frac{1}{E(N_{jj})}, j \in A \right)$$

Proof is by showing
that no limiting dist.
or $\pi = \pi P$ prob. soln.
can exist when chain is
transient or null rec;

hence must be pos. rec.

(See Lecture Notes
for proof)

If the state space \mathcal{S}
is finite ($|\mathcal{S}| < \infty$)
then an irreducible MC
is always pos. rec.

Proof: We already know
chain can only be either
positive or null recurrent
(transient not possible); so we

must show that

Null rec. is not

possible (See Lecture
Notes for Proof)

So, for irreducible finite state
MCS we can always try to
find π via $\pi = \pi P$

it always has a Prob. Soln.
 $\pi_j > 0, j \in \mathcal{A}$
 $\sum \pi_j = 1$

Examples: $\mathcal{S} = \{0, 1\}$ $P = \begin{pmatrix} .4 & .6 \\ .7 & .3 \end{pmatrix}$

$$\pi = (\pi_0, \pi_1)$$
$$\pi = \pi P$$

$$\pi_0 = .4\pi_0 + .7\pi_1$$
$$\pi_1 = .6\pi_0 + .3\pi_1$$
$$\pi_0 + \pi_1 = 1$$

$$\pi_0 = \frac{7}{13}, \pi_1 = \frac{6}{13}$$

$$\pi_1 = \frac{.6\pi_0}{.7}$$
$$\pi_0 + \pi_1 = 1$$
$$\pi_0 \left(1 + \frac{.6}{.7}\right) = 1$$

$$\delta = \{\sigma_j\}$$

$$0 < \alpha < 1$$

$$0 < \beta < 1$$

$$P = \begin{bmatrix} \alpha & 1-\alpha \\ \beta & 1-\beta \end{bmatrix}$$

$$\pi = \pi P$$

$$\pi_0 = \alpha \pi_0 + \beta \pi_1 \Rightarrow \pi_1 = \frac{(1-\alpha)\pi_0}{\beta}$$

$$\pi_1 = (1-\alpha)\pi_0 + (1-\beta)\pi_1$$

$$\pi_0 \left(1 + \frac{1-\alpha}{\beta}\right) = 1$$

$$\pi_0 + \pi_1 = 1$$

$$\pi_1 = \frac{1-\alpha}{\beta + 1-\alpha}$$

$$\pi_0 = \frac{1}{1 + \frac{1-\alpha}{\beta}} = \frac{\beta}{\beta + 1-\alpha}$$

$$S = \{0, 1, 2\}$$

$$P = \begin{bmatrix} .5 & .4 & .1 \\ .3 & .4 & .3 \\ .2 & .3 & .5 \end{bmatrix}$$

$$\pi = \pi P$$

Solvi

$$\pi = \left(\frac{21}{62}, \frac{23}{62}, \frac{18}{62} \right)$$

$$\pi_0 = .5\pi_0 + .3\pi_1 + .2\pi_2$$

$$\pi_1 = .4\pi_0 + .4\pi_1 + .3\pi_2$$

$$\pi_2 = .1\pi_0 + .3\pi_1 + .5\pi_2$$

$$\pi_0 + \pi_1 + \pi_2 = 1$$

$$E(T_{j0}) = \frac{1}{\pi_j}$$

$$= \frac{62}{21}$$

When a POS, rec. MC (X_n)
has been solved for π
we then set for free,

$$E(\tau_{jj}) = \frac{1}{\pi_j} \quad , j \in d$$

Expected
return times

$$\left(\pi_j \stackrel{\text{via}}{=} \frac{1}{E(\tau_{jj})} \right)$$

Quick proof that
the Simple Symmetric RW
($p = \frac{1}{2}$)

is Null versus pos.
recurrent

Suppose it was pos. rec.
then π exists (Prob. limiting dist.)

and $\pi_j = \frac{1}{E(\tau_{jj})} \geq 0 \quad j \in A$
 $\sum_j \pi_j = 1$

impossible because
for each $j \in \mathbb{Z}$

π_j has the same
dist. (hence same mean
ECT_j)
a constant

$$\Rightarrow \pi_j \equiv c > 0 \quad j \in \mathbb{Z}$$

$$\Rightarrow 1 = \sum_{j=-\infty}^{\infty} \pi_j = \sum_{j=-\infty}^{\infty} c = \infty \quad \text{(contradiction.)}$$

Null recurrence

Example of a pos. rec.
MC with infinite state space

Take the Simple Random walk restricted to stay in

$$\mathcal{S} = \{0, 1, 2, \dots\}$$

$$0 < p < 1$$

via $P_{0,0} = z, P_{0,1} = p$

$$P_{i,i+1} = p$$
$$P_{i,i-1} = z$$
$$i \geq 1$$

Clearly irreducible

let's try solving $\pi = \pi P$

$$P = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 & \dots & 4 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \end{matrix} & \begin{bmatrix} \varepsilon & p & 0 & 0 & \dots & 0 \\ \varepsilon & 0 & 0 & 0 & \dots & 0 \\ 0 & \varepsilon & 0 & p & \dots & 0 \\ 0 & 0 & \varepsilon & 0 & \dots & p \end{bmatrix} \end{matrix}$$

$$\begin{aligned} \pi_0 &= \varepsilon \pi_0 + \varepsilon \pi_1 & \Rightarrow & \quad p \pi_0 = \varepsilon \pi_1 \quad (1 - \varepsilon = p) \\ \pi_1 &= p \pi_0 + \varepsilon \pi_2 & \Rightarrow & \quad \pi_1 = \varepsilon \pi_1 + \varepsilon \pi_2 \quad \Rightarrow \end{aligned}$$

$$\boxed{p \pi_j = \sum \pi_{j+1}} \quad j \geq 0$$

$$\pi_1 = (p/\varepsilon) \pi_0$$

$$\pi_2 = (p/\varepsilon) \pi_1 = (p/\varepsilon)^2 \pi_0$$

$$\pi_j = \left(\frac{p}{\varepsilon}\right)^j \pi_0 \quad j \geq 0$$

$$\sum \pi_j = 1 \iff \pi_0 \sum_{j=0}^{\infty} \left(\frac{p}{\varepsilon}\right)^j = 1 \iff$$

$q > p$ so that

$$p/q < 1.$$

$$\pi_0 = 1 - p/q$$

$$\Rightarrow \pi_j = \left(\frac{p}{q}\right)^j (1 - p/q) \quad j \geq 0$$

(Geometric Series!)
 $\sum_{i=0}^{\infty} q^i = \frac{1}{1-q}$
for $|q| < 1$

Geometric dist. with mass

at 0, $\pi_0 > 0$

once we have solved Π
for a pos. rec. MC
we might wish to
compute

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=1}^N X_n = \sum_{i \in \mathcal{A}} j_i \Pi_i$$

long-run average = mean of the
limiting dist.

$$\frac{1}{N} \sum_{n=1}^N X_n = \sum_{j \in \mathcal{A}} j \frac{N_j(n)}{N}$$

Let $N_j(n) = \sum_{n=1}^N \mathbb{I}\{X_n = j\}$ $\xrightarrow{N \rightarrow \infty}$

$\lim_{N \rightarrow \infty} \frac{N_j(n)}{N} = \pi_j$ w.p.1 by def.

$$\sum_{j \in \mathcal{A}} j \pi_j$$

if (X_n) is ≥ 0 it holds

More Generally Suppose
for some function f

You want

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=1}^N f(X_n)$$

$$\left(\begin{array}{l} f(x) = x \\ f(x) = x^2 \\ f(x) = I\{|x| \leq a\} \\ \text{etc.} \end{array} \right)$$

$$= \sum_{j \in \mathcal{J}} f(j) \pi_j$$

$$\left(\begin{array}{l} \text{if } f \geq 0 \text{ or} \\ f \text{ is bdd or} \\ \sum_{j \in \mathcal{J}} |f(j)| \pi_j < \infty \end{array} \right)$$

Stationarity

if (X_n) is pos. rec.

with limiting dist. $\overline{\Pi} = (\overline{\pi}_j)$

and if the chain is started off initially distributed as $\overline{\Pi}$

$$P(X_0 = j) = \overline{\pi}_j \quad j \in \mathcal{I}$$

then for each $n \geq 1$, $P(X_n = j) = \overline{\pi}_j$
 $j \in \mathcal{I}$.

Proof:

Suppose $P(X_0 = i) = \pi_i, i \in \mathcal{A}$

$P(X_1 = j)$

$$= \sum_{i \in \mathcal{A}} P(X_1 = j | X_0 = i) \pi_i$$

$$= \sum_{i \in \mathcal{A}} \pi_i P_{ij} = \pi_j \quad \text{via } \pi = \pi P$$

Continuing by induction on $n \geq 1$

$$P(X_{n+1}=j) = \sum_{i \in \Delta} \underbrace{P(X_{n+1}=j | X_n=i)} \pi_i$$

$$= \sum_{i \in \Delta} \pi_i \cdot p_{ij} = \pi_j$$

X_n has same dist. π for all $n \geq 0$

If $X_0 \stackrel{\text{dist.}}{=} \pi$, then

$\{X_n\}$ is called a

Stationary version of the

MC.

$$\delta = (1, 2)$$

$$\pi = \left(\frac{1}{2}, \frac{1}{2}\right) \checkmark$$

$$P = \begin{matrix} 1 & 2 \\ \left[\begin{array}{cc} 0 & 1 \\ 1 & 0 \end{array} \right] \end{matrix}$$

$$\{X_n\} = \begin{cases} (1, 2, 1, 2, \dots) & \text{if } x_0 = 1 \\ (2, 1, 2, 1, \dots) & \text{if } x_0 = 2 \end{cases}$$

Let $\{X_n^*\}$ denote a stationary version

$$\{X_n^*\} = \left\{ \begin{array}{l} \{ \underset{\downarrow}{1}, \underset{\downarrow}{2}, \underset{\downarrow}{2}, \dots \} \quad \text{wp} = \frac{1}{2} \\ \{ \underset{\downarrow}{2}, \underset{\downarrow}{1}, \underset{\downarrow}{2}, \underset{\downarrow}{1}, \dots \} \quad \text{wp} = \frac{1}{2} \end{array} \right.$$

time n : 0 1 2 3

$$P(X_n^* = 1) = \frac{1}{2} = P(X_n^* = 2) \quad \text{for all } n$$