

DISTRIBUTION of MARKOV CHAIN

X_0, X_1, X_2, \dots discrete time M.C.

$X_n \in S$ finite or countable state space

$P = (p_{xy})_{x,y \in S}$ STOCHASTIC MATRIX

$v^{(0)}$ initial distribution : $v_x^{(0)} = P(X_0 = x)$

$v^{(n)}$ distribution at time n $v_x^{(n)} = P(X_n = x)$

$v^{(n)} = \mu_{X_n}$

$v^{(n)} \in \mathbb{R}^{|S|}$ if S FINITE, $v^{(n)} \in \mathbb{R}^\infty$ if $|S| = +\infty$

distribution $v^{(n)}$ has properties :

- $v_x^{(n)} \geq 0$

$$\sum_{x \in S} v_x^{(n)} = 1$$

How to compute $v^{(n)}$ knowing $v^{(0)}$ and P

$$v_x^{(1)} = P(X_1 = x) = \sum_y P(X_1 = x | X_0 = y) P(X_0 = y)$$

\downarrow
FORMULA TOTAL PROBABILITY

$$= \sum_{y \in S} p_{yx} v_y^{(0)}$$

num can be infinite

$$\Rightarrow v^{(1)} = v^{(0)} P$$

"matrix product", DISTRIBUTIONS as ROW VECTORS

In the same way, we find

$$v^{(n+1)} = v^{(n)} P \quad \forall n$$

From this, we can compute the distribution at time n :

$$v^{(n)} = v^{(n-1)} P = (v^{(n-2)} P) \cdot P = \dots$$

$$= v^{(0)} P^n$$

Denote $p_{xy}^n = P(X_n = y | X_0 = x)$

$$\Rightarrow p_{xy}^n = (P^n)_{xy}$$

\hookrightarrow result from the n -PRODUCT P^n

$$\Rightarrow p_{xy} = (P)_{xy}$$

\hookrightarrow component (x,y) of n -PRODUCT P^n

$$[\text{note } p_{xy}^n \neq (p_{xy})^n]$$

Product of matrices usually done if $|S| < \infty$

Otherwise, we prove using MARKOV PROPERTY that

$$v^{(n+2)} = v^{(n)} P^2$$

$$\text{where } (P^2)_{yx} = \sum_{z \in S} p_{yz} p_{zx}$$

- $P(X_{n+2} = x | X_n = y) = p_{yx}^2$

$$\text{TOTAL PROB.} = \sum_{z \in S} P(X_{n+2} = x, X_{n+1} = z | X_n = y)$$

$$P(A \cap B | C)$$

$$P(A | B \cap C) \cdot P(B | C) = \sum_{z \in S} P(X_{n+2} = x | X_{n+1} = z, X_n \neq y) P(X_{n+1} = z | X_n = y)$$

$$\text{MARKOV} \Rightarrow = \sum_{z \in S} P(X_{n+2} = x | X_{n+1} = z) P(X_{n+1} = z | X_n = y)$$

$$= \sum_{z \in S} p_{zx} p_{yz}$$

$$\Rightarrow p_{yx}^2 = (P^2)_{yx}$$

- $p_{yx}^n = P(X_{m+n} = x | X_m = y) \quad \forall n, \forall m$

$$= (P^n)_{yx}$$

$$v^{(m+n)} = v^{(m)} P^n \quad \forall n, m$$

\rightarrow CHAPMAN - KOLMOGOROV EQUATION

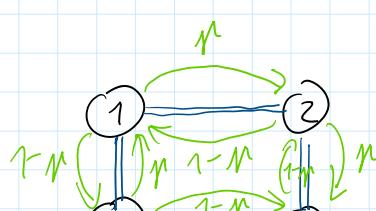
$$P^{n+m} = P^n \cdot P^m$$

$$\text{where } (P^n)_{xy} = p_{xy}^n$$

Exemple (SENTINEL)

$$S = \{1, 2, 3, 4\}$$

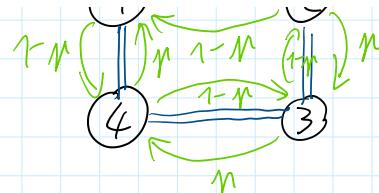
TOWERS \leftrightarrow CASTLE



1 1 - 1 1 1

TOWERS of CASTLE

SENTINEL moves



X_n = position of sentinel at time n

SENTINEL moves to NEXT TOWER prob. p

PREVIOUS TOWER (CLOCKWISE ORDER) prob. $1-p$

$$p_{12} = p, p_{14} = 1-p, p_{13} = 0, p_{11} = 0$$

$$P = \begin{pmatrix} 0 & p & 0 & 1-p \\ 1-p & 0 & p & 0 \\ 0 & 1-p & 0 & p \\ p & 0 & 1-p & 0 \end{pmatrix}$$

Note 2 steps to return to the same state

(with positive probability)

EVEN numbers have property that 2 steps are required to go to another even number

CLASSIFY THE STATES

Note that a MARKOV CHAIN can be represented by

a DIRECTED GRAPH

$$\text{GRAPH} = \{ \text{VERTEX}, \text{EDGES} \}$$



$$\text{VERTEX} = \text{STATES} = S$$

\exists EDG FROM STATE x to y is $p_{xy} > 0$

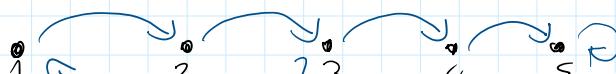
COMMUNICATION

We say that $j \in S$ is ACCESSIBLE from $i \in S$ if

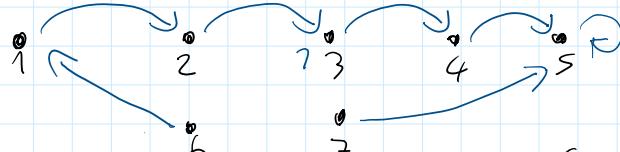
$$\exists n > 0 \text{ s.t. } p_{ij}^n > 0$$

$$[p_{ii}^0 = 1]$$

example



example

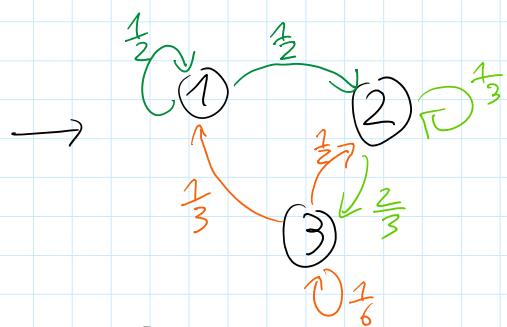


- 5 is ACCESSIBLE from 1 (in 4 steps) ($p_{15}^{(4)} > 0$)
- 1 is NOT ACCESSIBLE FROM 3

- We say that i and j COMMUNICATE if i is accessible from j and j is accessible from i
- COMMUNICATION is an EQUIVALENT RELATION
 - we can divide the state space in equivalent classes COMMUNICATION CLASSES
 - elements in same communication class share the same properties (PERIODICITY, RECURRENCE, ...)

Example

$$P = \begin{pmatrix} 1 & \frac{1}{2} & 0 \\ \frac{1}{2} & \frac{1}{2} & \frac{2}{3} \\ 0 & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{3} & \frac{1}{2} & \frac{1}{6} \end{pmatrix}$$



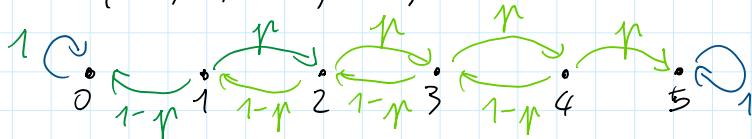
Just one COMMUNICATION CLASS $\{1, 2, 3\}$

We say that the MARKOV CHAIN IS IRRREDUCIBLE if all states communicate with each other (just one communication class)

Exercise MC SENTINEL is also irreducible

Example GAMBLER'S RUIN MODEL

$$S = \{0, 1, \dots, m\}$$



Communication classes? 3

$$C_1 = \{0\}$$

$$C_2 = \{5\}$$

$$C_3 = \{1, 2, 3, 4\}$$

When $C = \{i\}$, i is called ABSORBING STATE

STATIONARY DISTRIBUTION

INVARIANT or STATIONARY distribution π

if does not change in time,

i.e. $X_0 \sim \pi \Rightarrow X_1 \sim \pi$

- Questions
- existence / uniqueness of invariant distribution?
 - properties of π ?
 - Convergence to the invariant distribution?

$\nu^{(0)} = \pi$, we look for π s.t. $\nu^{(1)} = \pi$

$$\pi = \nu^{(1)} = \nu^{(0)} P = \pi P$$

- INVARIANT DISTRIBUTION is SOLUTION to equation

$$\pi = \pi P$$

$$\text{and } \pi_i \geq 0 \quad \forall i \in S \quad \text{and} \quad \sum_{i \in S} \pi_i = 1$$

$\pi(P - I) = 0 \rightarrow \pi$ is LEFT EIGENVECTOR of eigenvalue 1 (if $|S| < \infty$)

Thm 1 if S is FINITE and P is IRREDUCIBLE

Then there exists a unique invariant distribution π
AND π is such that $\pi_x > 0 \quad \forall x$

Exercise $|S| = 3$

$$P = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{3} & \frac{1}{2} & \frac{1}{6} \end{pmatrix} \quad \text{is IRREDUCIBLE}$$

Compute π invariant distribution

$$\begin{cases} \pi = \pi P \\ \pi_1 + \pi_2 + \pi_3 = 1 \end{cases}$$

$$(\pi_1, \pi_2, \pi_3) = (\pi_1, \pi_2, \pi_3) \begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{3} & \frac{1}{2} & \frac{1}{6} \end{pmatrix}$$

$$\begin{cases} \pi_1 = \frac{\pi_1}{2} + \frac{\pi_3}{3} \\ \pi_2 = \frac{\pi_1}{2} + \frac{\pi_2}{3} + \frac{\pi_3}{2} \\ \pi_3 = \frac{2}{3}\pi_2 + \frac{\pi_3}{6} \\ \pi_1 + \pi_2 + \pi_3 = 1 \end{cases}$$

$$\Rightarrow \pi = \left(\frac{8}{35}, \frac{15}{35}, \frac{12}{35} \right)$$

• Convergence? given $v^{(0)}$

is it true that $\lim_{n \rightarrow \infty} v^{(n)} = \pi$ ERGODIC THEOREM

true if NC is APERIODIC (and $|S| < \infty$)