Stochastíc Processes and Markov Chains II



(1856-1922)

Markov Process

(a) *States* S₁, S₂,..., S_r and Steps

(b) *Transition Probabilities pij* = probability of moving from State *i* to State *j* in a single step

 p_{ij} = Pr[S_j at step(n+1) | S_i at step n]

Note: Each $p_{ij} \ge 0$

$$\sum_{j=1}^{r} p_{ij} = 1 \text{ for all } i$$

 p_{ij} 's are independent of n

(c) Initial Distribution $\overline{p^{(o)}}$

Two Special Types of Markov Process

REGULAR

Some Power of the Transition Matrix Has All Positive Entries

ABSORBING

Social Mobility Example A representation of transition matrix **P**

		generation $(n+1)$		[children]	
		Р	\mathbf{S}	U	
generation n	Р	.8	.1	.1	
[parents]	\mathbf{S}	.2	.6	.2	
	U	.25	.25	.5	

An Absorbing Matrix						
		Α	B	C		
step n	Α	1	0	0		
	B	.2	.6	.2		
	C	0	. 5	.5		

A is an *Absorbing* State

B, C are Transient States

What is the Long Term Behavior?

$$\lim_{n\to\infty} \vec{p}^{(n)} = \vec{p}^{(0)} \lim_{n\to\infty} P^n$$

$$P = \begin{pmatrix} .8 & .1 & .1 \\ .2 & .6 & .2 \\ .25 & .25 & .5 \end{pmatrix}$$

Run the program MARKOV to test:

 $\vec{p}^{(n)}$ seems to approach (.53, .26, .21)

Can we find $\vec{w} = \lim_{n \to \infty} \vec{p}^{(n)}$ analytically?

 $\vec{w} = \vec{w}P$ where $\vec{w} = (x, y, z)$ and x + y + z = 1

$$(x, y, z) = (x, y, z) \begin{pmatrix} .8 & .1 & .1 \\ .2 & .6 & .2 \\ .25 & .25 & .5 \end{pmatrix}$$

$$\begin{cases} x = .8x + .2y + .25z \\ y = .1x + .6y + .25z \\ z = .1x + .2y + .5z \end{cases}$$

(1)
$$-.2x + .2y + .25z = 0$$

(2) $1x - .4y + .25z = 0$
(3) $.1x + .2y - .5z = 0$

Rewriting (1) and (2):

.25z = .2x - .2y $.25z = -.1x + .4y \implies .2x - .2y = -.1x + .4y$

$$\Rightarrow .3x = .6y \Rightarrow y = \frac{1}{2}x$$

Substitute in (3): $1x + .2(.5x) - .5z = 0 \Rightarrow .2x = .5z \Rightarrow z = \frac{2}{5}x$

Then
$$x + y + z = 1$$

becomes $x + \frac{1}{2}x + \frac{2}{5}x = 1$
 $10x + 5x + 4x = 1C$
 $19x = 1C$
 $x = 10/19$
 $y = 5/19$
 $z = (2/5)(10/19) = 4/19$

long term: $\left(\frac{10}{19}, \frac{5}{19}, \frac{4}{19}\right) \approx (.526, .263, .211)$

Main Result on Regular Markov Chains and Examples

Definition. A Markov Process is **regular** if some positive power of the transition matrix has all positive entries.

Example.

$$P = \begin{pmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix} \Longrightarrow P^{2} = \begin{pmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{4} & \frac{3}{4} \end{pmatrix}$$

Theorem: If P is the transition matrix of a regular Markov Process, then there is a vector **w** all of whose components are strictly positive such that

(1)
$$\mathbf{w} = \lim_{n \to \infty} \mathbf{p}^{(n)}$$

(2) w is independent of $\mathbf{p}^{(0)}$

(3) w is the unique probability vector satisfying w = w P

(4)
$$\lim_{n \to \infty} P^n = \begin{pmatrix} \mathbf{w} \\ \mathbf{w} \\ \dots \\ \dots \\ \mathbf{w} \end{pmatrix}$$

(5) w is a left eigenvector of P associated with eigenvalue 1

$$\begin{aligned} Example : \text{If } \mathbf{w} &= (x, y), \text{ then} \\ (x, y) \begin{pmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix} = (x, y) \Rightarrow \begin{array}{l} x = 0x + \frac{1}{2}y \\ y = 1x + \frac{1}{2}y \end{array} \Rightarrow \begin{array}{l} y = 2x \\ y = 2x \\ y = 2x \\ \text{but } x + y = 1 \text{ so } 1 - x = 2x \text{ so } x = 1/3 \\ \text{and } \mathbf{w} &= \left(\frac{1}{3}, \frac{2}{3}\right) \end{aligned}$$

$$Check: \left(\frac{1}{3}, \frac{2}{3}\right) \begin{pmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix} = \left(\frac{1}{3}, \frac{1}{3} + \frac{1}{3}\right) \end{aligned}$$

A Baseball Example



Can check by Induction that $\mathbf{p}^{(n)} = \left(\frac{3}{4} + \left(\frac{-1}{5}\right)^n \left(p - \frac{3}{4}\right), \frac{1}{4} - \left(\frac{-1}{5}\right)^n \left(p - \frac{3}{4}\right)\right)$ so $\lim_{n \to \infty} \mathbf{p}^{(n)} = \left(\frac{3}{4}, \frac{1}{4}\right)$

Solving by $\mathbf{w} = \mathbf{w}P$ gives $(x, y) \begin{pmatrix} .7 & .3 \\ .9 & .1 \end{pmatrix} = (x, y)$ or (x, y) = (.7x + .9y, .3x + .1y) $x = .7x + .9y \Rightarrow -.3x = .9y \Rightarrow y = \frac{x}{3}$ but x + y = 1 so $x + \frac{x}{3} = 1$ or $x = \frac{3}{4}$ Eigenvalue/Eigenvector Analysis

$$P = \begin{bmatrix} \frac{7}{10} & \frac{3}{10} \\ \frac{9}{10} & \frac{1}{10} \end{bmatrix} \Rightarrow P - \lambda I = \begin{bmatrix} \frac{7}{10} - \lambda & \frac{3}{10} \\ \frac{9}{10} & \frac{1}{10} - \lambda \end{bmatrix}$$

so
$$\det(P - \lambda I) = (\frac{7}{10} - \lambda)(\frac{1}{10} - \lambda) - (\frac{9}{10})(\frac{3}{10})$$

and $\det(P - \lambda I) = 0$ becomes
 $\lambda^2 - \frac{8}{10}\lambda - \frac{20}{100} = 0$
 $\Rightarrow 5\lambda^2 - 4\lambda - 1 = 0 \Rightarrow (5\lambda + 1)(\lambda - 1) = 0$ so $\lambda = -\frac{1}{5}$ or $\lambda = 1$

Since there are 2 distinct eigenvalues, there is a 2 x 2 invertible matrix S such that

$$S^{-1}PS = \begin{bmatrix} 1 & 0 \\ 0 & \frac{-1}{5} \end{bmatrix} \text{ so that } P = S \begin{bmatrix} 1 & 0 \\ 0 & \frac{-1}{5} \end{bmatrix} S^{-1} \text{ and hence}$$
$$P^{n} = S \begin{bmatrix} 1 & 0 \\ 0 & \frac{-1}{5} \end{bmatrix}^{n} S^{-1} = S \begin{bmatrix} 1^{n} & 0 \\ 0 & (\frac{-1}{5})^{n} \end{bmatrix} S^{-1} = S \begin{bmatrix} 1 & 0 \\ 0 & (\frac{-1}{5})^{n} \end{bmatrix} S^{-1}$$

The matrix S can be formed by taking right eigenvectors associated with the eigenvalues as its columns. An eigenvector associated with $\lambda = 1$ is $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$ and an eigenvector associated with $\lambda = -\frac{1}{5}$ is $\begin{pmatrix} 1 \\ -3 \end{pmatrix}$. Thus we can let $S = \begin{bmatrix} 1 & 1 \\ 1 & -3 \end{bmatrix}$. Then $S^{-1} = \begin{bmatrix} \frac{3}{4} & \frac{1}{4} \\ \frac{1}{4} & -\frac{1}{4} \end{bmatrix}$.

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Thus
$$P^{n} = \begin{bmatrix} 1 & 1 \\ 1 & -3 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & (\frac{-1}{5})^{n} \end{bmatrix} \begin{bmatrix} \frac{3}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{-1}{4} \end{bmatrix}^{=}$$
$$\begin{bmatrix} 1 & (\frac{-1}{5})^{n} \\ 1 & -3(\frac{-1}{5})^{n} \end{bmatrix} \begin{bmatrix} \frac{3}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{-1}{4} \end{bmatrix}$$
so we have
$$P^{n} = \begin{bmatrix} \frac{3}{4} + \frac{1}{4}(\frac{-1}{5})^{n} & \frac{1}{4} - \frac{1}{4}(\frac{-1}{5})^{n} \\ \frac{3}{4} - \frac{3}{4}(\frac{-1}{5})^{n} & \frac{1}{4} + \frac{3}{4}(\frac{-1}{5})^{n} \end{bmatrix}$$