

CONSIDER $R=2$. (2 action case)

$$p_1(n+1) = (1-a) p_1$$

$$\text{if } d = d_2 \quad \beta = 0$$

$$= 1 - p_2 + b p_2$$

$$\text{if } d = d_2 \quad \beta = 1$$

$$= 1 - (1-a) p_2$$

$$\text{if } d = d_1 \quad \beta = 0$$

$$= (1-b) p_1$$

$$\text{if } d = d_1 \quad \beta = 1$$

GENERAL $L_{R,P}$ (LINEAR REWARD. PENALTY SCHEME)

$$d = d_2, \beta = 1$$

$$p_2 \leftarrow (1-b) p_2$$

$$\therefore p_1(n+1) = 1 - (1-b) p_2 \\ = 1 - p_2 + b p_2.$$

$$d = d_1, \beta = 0$$

$$p_2 \leftarrow (1-a) p_2$$

$$\therefore p_1 \leftarrow 1 - p_2 = 1 - (1-a) p_2$$

SIMPLE CASE. $a = b$. (Symmetric $L_{R,P}$ Scheme).

$$\begin{aligned} p_i(n+1) &= k p_1 && \text{if } d = d_1, \beta = 1 \\ &= 1 - k p_2 && \text{if } d = d_1, \beta = 0 \\ &= 1 - k p_2 && \text{if } d = d_2, \beta = 1 \\ &= k p_1 && \text{if } d = d_2, \beta = 0 \end{aligned}$$

NOTE: $(d_2, \beta = 0) \equiv (d_1, \beta = 1)$

AND. $(d_1, \beta = 0) \equiv (d_2, \beta = 1)$

NOTE

p_1, p_2 are both RANDOM VARIABLES

Previously: $\underline{p}(n) = F^T p(n-1)$ (PSSA)

NOW WE CAN'T DO THAT.

$p_i(n+1)$ can take 4 values - depending on outcome.

i.e. $p_i(n+1)$ is a r.v. with a mean, variance etc.

WHAT IS $E[p_i(n+1)]$? USE ONLY p_i in Expression

$$\begin{aligned} p_i(n+1) &= k p_i && \text{w.p. } p_i c_1 \\ &= 1 - k(1 - p_i) && \text{w.p. } p_i(1 - c_1) \\ &= 1 - k(1 - p_i) && \text{w.p. } (1 - p_i) c_2 \\ &= k p_i && \text{w.p. } (1 - p_i)(1 - c_2) \end{aligned}$$

$\therefore E[p_i(n+1)]$ is a function of p_i

$$\begin{aligned} E[p_i(n+1) | p_i] &= k p_i (p_i c_1 + (1 - p_i)(1 - c_2)) \\ &\quad + (1 - k(1 - p_i)) (p_i(1 - c_1) + (1 - p_i) c_2) \end{aligned}$$

NOTE: All p_i^2 terms luckily cancel.

$$E[p_i(n+1) | p_i] = p_i - k(c_1 + c_2)p_i + k c_2$$

$$E[p_i(n+1)] = [1 - k(c_1 + c_2)] E[p_i] + k c_2$$

a linear difference equation...

$$y(n+1) = e y(n) + f$$

General Solution.

In THE LIMIT. $p_i(n+1) = p_i(n)$

$$\bar{p}_i(n+1) = [1 - k(c_1 + c_2)] \bar{p}_i(n) + k c_2$$

is.

$$\bar{p}_i(\infty) = [1 - k(c_1 + c_2)] \bar{p}_i(\infty) + k c_2$$

$$k(c_1 + c_2) \bar{p}_i(\infty) = k c_2$$

or $E[\bar{p}_i(\infty)] = c_2 / (c_1 + c_2)$

THEOREM.

1. The Symmetric $L_{R,p}$ scheme has

$$E[\bar{p}_i(\infty)] = \frac{c_j}{c_i + c_j} \quad i \neq j, \quad i=1,2$$

AND IS THEREFORE EXPEDIENT.

2. THE LIMITING EXP. VALUE OF $\bar{p}_i(\infty)$ IS INDEPENDENT OF THE PARAMETER OF $L_{R,p}$ scheme.

IS THERE A "BETTER" WAY TO LOOK AT THIS?

YES!

$$E[p_1(n+1) | p_1] = p_1 - k(C_1 + C_2)p_1 + kC_2.$$

i.e.: writing it in terms of both p_1 & p_2 .

$$(kC_2 = (p_1 + p_2)(kC_2))$$

$$E[p_1(n+1) | p_1] = p_1 [1 - k(C_1 + C_2) + kC_2] + kC_2 p_2$$

$$= p_1 (1 - kC_1) + kC_2 p_2.$$

$$\therefore E[p_1(n+1)] = (1 - kC_1) E[p_1] + kC_2 E[p_2]$$

$$E[p_2(n+1)] = kC_1 E[p_1] + (1 - kC_2) E[p_2]$$

$$E \begin{bmatrix} p_1(n+1) \\ p_2(n+1) \end{bmatrix} = \begin{bmatrix} 1 - kC_1 & kC_1 \\ kC_2 & 1 - kC_2 \end{bmatrix}^T E \begin{bmatrix} p_1 \\ p_2 \end{bmatrix}$$

i.e. $E \begin{bmatrix} p_1 \\ p_2 \end{bmatrix}$ are the "STATES OF AN ERGODIC M.C."

$$F = \begin{bmatrix} 1 - kC_1 & kC_1 \\ kC_2 & 1 - kC_2 \end{bmatrix}$$

Markov matrix

SOLUTION is $\begin{bmatrix} \frac{C_2}{C_1 + C_2} \\ \frac{C_1}{C_1 + C_2} \end{bmatrix}$ just as we thought...

i.e. $E[p_i(\infty)] \rightarrow \frac{C_i}{C_1 + C_2}$.

EIGENVALUES : ALWAYS ONE IS UNITY.

OTHER : $(1 - kC_1 + 1 - kC_2) = 1 + \lambda_2$

$$\lambda_2 = 1 - k(C_1 + C_2)$$

RATE OF CONVERGENCE determined by 'k'

VARIANCE

$$\text{Var } p_i(\infty) = \frac{C_1 C_2}{(C_1 + C_2)^2} \cdot \frac{(1 - k)^2}{(1 - k)^2 + 2k(1 - k)(C_1 + C_2)}$$