

## Extortion strategies & Zero Determinant (ZD) strategies

Here, we return to the iterated PD models with  $\begin{pmatrix} R & S \\ T & P \end{pmatrix}$  as usual, e.g.  $\begin{pmatrix} 3 & 0 \\ 5 & 1 \end{pmatrix}$

Press + Dyson show first, an interesting result. If a short memory player plays a long memory player, the short memory players score is the same as if the long memory guy played a short memory strategy. Because of this result, we can derive strategies for X (short mem) assuming each player has memory 1!

Let  $X, Y$  be random variables with values  $x, y$  that are the players respective moves on a given iteration. Suppose player X keeps history  $H_0$  but Y keeps longer history  $H_1$ .

We want to show that the joint prob distribution of  $(x, y)$  given history  $(H_0, H_1)$

averaged over  $(H_0, H_1)$  is same as joint prob of  $(x, y)$  averaged over shorter history  $H_0$

$$\begin{aligned} \langle P(x, y | H_0, H_1) \rangle_{H_0, H_1} &\stackrel{\Delta}{=} \sum_{H_0, H_1} P(x, y | H_0, H_1) P(H_0, H_1) \\ &= \sum_{H_0, H_1} P(x | H_0) P(y | H_0, H_1) P(H_0, H_1) \xleftarrow[\text{indep}]{\text{conditional ??}} \\ &= \sum_{H_0} P(x | H_0) \left[ \sum_{H_1} P(y | H_0, H_1) P(H_1 | H_0) P(H_0) \right] \xrightarrow[\text{sum over } H_1]{\text{definition of conditional}} \\ &= \sum_{H_0} P(x | H_0) \left[ \sum_{H_1} P(y | H_1 | H_0) \right] P(H_0) \xleftarrow[\text{give } H_1]{\text{marginal}} \\ &= \sum_{H_0} P(x | H_0) P(y | H_0) P(H_0) \\ &= \langle P(x, y | H_0) \rangle_{H_0} \end{aligned}$$

Intuitively - from X's point of view, X views Y's long strategy as a peculiar random number generator. Thus the player with the shortest memory sets the rules of the game.

ZD strategies:

As usual let  $x, y \in \{cc, cd, dc, dd\}$ , where c, d are cooperate, defect  
 X's strategy is  $\vec{p} = (p_1, p_2, p_3, p_4)$  are probabilities to cooperate given the  
 outcome +  $\vec{q} = (q_1, q_2, q_3, q_4)$  are Y's strategies, seen from his per-  
 spective, i.e.  $y|x = \{cc, cd, dc, dd\}$  (Note this is a little diff than we  
 used before). So payoff of Y

$$\begin{bmatrix} p_i q_j & p_i (1-q_j) \\ (1-p_i) q_j & (1-p_i)(1-q_j) \end{bmatrix}$$

As before we get a markov transition matrix:

$$M = \begin{bmatrix} p_1 q_1 & p_1 (1-q_1) & (1-p_1) q_1 & (1-p_1)(1-q_1) \\ p_1 q_3 & p_1 (1-q_3) & (1-p_1) q_3 & (1-p_1)(1-q_3) \\ p_3 q_2 & p_3 (1-q_2) & (1-p_3) q_2 & (1-p_3)(1-q_2) \\ p_4 q_4 & p_4 (1-q_4) & (1-p_4) q_4 & (1-p_4)(1-q_4) \end{bmatrix}$$

$\vec{V}$  steady state (left eigenvector):  $\vec{V}^T M = \vec{V}^T \Rightarrow \vec{V}^T M' = 0$

$$M' = M - I \Rightarrow \det M' = 0$$

Recall cramer's rule:  $\text{Adj}(M') M' = \det(M') I = 0$   
 $(4 \times 4) \begin{pmatrix} a_{11} & \dots & a_{14} \\ \vdots & \ddots & \vdots \\ a_{41} & \dots & a_{44} \end{pmatrix} M' = 0$

$\text{Adj}(M') = \text{matrix of minors } (3 \times 3) \text{ determinants, since } V^T M' = 0$

This means every row of  $\text{Adj}(M')$  is proportional to  $\vec{V}$

$$\begin{pmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \\ m_{41} & m_{42} & m_{43} & m_{44} \end{pmatrix} C_{ij} \text{ cofactor } \text{Adj}(M') = C^T$$

so, for example, up to sign, the components of  $\vec{V}$  are the fourth row  
 of  $\text{Adj}(M')$ , which are the cofactors of the 4th column of  $M'$   
 which are just the determinants of the first three columns of  $M'$

leaving out the relevant row. These determinants are unchanged  
 by adding the first column of  $M'$  to the second + third columns

Let  $\vec{f}$  be arbitrary vector. Then  $\vec{V} \cdot \vec{f}$  is easy to compute using the above manipulation:

$$\vec{V} \cdot \vec{f} = \det \begin{bmatrix} -1 + p_1 q_1 & -1 + p_1 & -1 + q_1 & f_1 \\ p_2 q_2 & -1 + p_2 & q_2 & f_2 \\ p_3 q_3 & p_3 & -1 + q_3 & f_3 \\ p_4 q_4 & p_4 & q_4 & f_4 \end{bmatrix} = D(p, q, f)$$

$$= \tilde{p} \quad \tilde{q}$$

Note that two columns depend only on one player's strategy!

$$\vec{S}_X = [R, S, T, P], \vec{S}_Y = [R, T, S, P]$$

$$s_X = \frac{V \cdot \vec{S}_X}{V \cdot \vec{1}} = \frac{D(p, q, \vec{S}_X)}{D(p, q, \vec{1})}, s_Y = \frac{V \cdot \vec{S}_Y}{V \cdot \vec{1}} = \frac{D(p, q, \vec{S}_Y)}{D(p, q, \vec{1})}$$

$$\alpha s_X + \beta s_Y + \gamma = \frac{D(p, q, \alpha \vec{S}_X + \beta \vec{S}_Y + \gamma \vec{1})}{D(p, q, \vec{1})} \quad \text{Really cool!}$$

If I choose  $\hat{P} = R[\alpha \vec{S}_X + \beta \vec{S}_Y + \gamma \vec{1}]$ , then this determinant vanishes since two columns are proportional and this means I enforce a linear relationship between the payoffs no matter what  $q$ !!!

If course,  $\hat{P}_Y$  may not be feasible since  $\hat{P}_i \in [0, 1]$

EXAMPLE X sets Y's score!

Set  $\alpha = 0$ . Use  $\hat{P} = \beta \vec{S}_Y + \gamma \vec{1}$  & solve for  $p_2, p_3$  in terms of  $p_1, p_4$

eliminating  $\beta, \gamma$ . Show:

$$p_2 = \frac{p_1(T-P) - (1+p_4)(T-R)}{R-P}, p_3 = \frac{(1-p_1)(P-S) + p_4(R-S)}{R-P}$$

$$\text{To find } s_Y = \frac{(1-p_1)P + p_4 R}{(1-p_1) + p_4}, \text{ in PD } T > R > P > S$$

so there is a feasible strategy when  $p_1 \approx 1, p_4 \geq 0 \Rightarrow p_2 \approx 1,$

$p_3 \approx 0$ . Clearly  $s_Y$  is weighted average of  $1-p_1 + p_4$  so all possible scores between  $P+R$  are possibly for Y. X can completely ignore Y but set Y's score. X can shoot Y & then play a better strategy

X cannot really set his own score:  $\tilde{p} = \alpha \vec{s}_x + \gamma \mathbb{1}$

$$p_2 = \frac{(1+\phi)(R-S) - p_1(P-S)}{R-P} \geq 1, \quad p_3 = \frac{-(1-p_1)(T-P) - p_4(T-R)}{R-P} \leq 0$$

or never coop

Only one feasible point:  $(1, 1, 0, 0)$  which is always cooperate.  $S \cup$

X cannot unilaterally set his score.

"Extortion" Suppose  $\tilde{p} = \phi [(\vec{s}_x - P\mathbb{1}) - \chi(\vec{s}_y - P\mathbb{1})]$ . Then this

$$\text{means } \phi[(s_x - P) - \chi(s_y - P)] = 0 \Rightarrow (s_x - P) = \chi(s_y - P) \Rightarrow$$

X can make his gain above mutual detection ( $P$ )  $\chi$  times greater than Y's !!

Solving the equations for  $p_i$ :

$$p_1 = 1 - \phi(\chi - 1) \frac{R-P}{P-S}, \quad p_2 = 1 - \phi(1 + \chi \frac{T-P}{P-S}), \quad p_3 = \phi(\chi + \frac{T-P}{P-S}), \quad p_4 = 0$$

so feasible strategies exist for any  $\chi > 0$  and small  $\phi$ , eg

$$0 \leq \phi \leq \frac{(P-S)}{(P-S) + \chi(T-P)}$$

Clearly X's score depends on Y's strategy and both are maximized when Y cooperates:  $(1, 1, 1, 1)$  in which case

$$s_x = \frac{P(T-R) + \chi[R(T-S) - P(T-R)]}{(T-R) + \chi(R-S)} \quad (\chi=1 \Rightarrow s_x = s_y = R)$$

E.G.  $(5, 3, 1, 0)$ ,

$$\tilde{p} = [1 - 2\phi(\chi - 1), 1 - \phi(4\chi + 1), \phi(\chi + 4), 0], \quad \text{ok for } 0 \leq \phi \leq (4\chi + 1)^{-1}$$

$$s_x = \frac{2+3\chi}{2+3\chi}, \quad s_y = \frac{12+3\chi}{2+3\chi}. \quad \text{Mutual 100n} \Rightarrow 3 \quad s_x > 3 + \chi > 1$$

$s_y < 3$  !!.

As  $\chi \rightarrow \infty$ ,  $s_x \rightarrow \frac{13}{3} + \chi \rightarrow 1$  so Y has no reason to cooperate  
so X should not get too greedy!

Say  $\chi = 3 + \phi$  midpoint of feasibility  $\vec{p} = \left(\frac{11}{13}, \frac{1}{2}, \frac{7}{26}, 0\right)$

$s_x \approx 3.73, s_y \approx 1.91$ . Note  $\chi=1, \phi=1/5 \rightarrow \text{TFT}$

$(1, 0, 1, 0)$

## Spatial game in a continuum

This is my version of a spatial game. We will later look at some others. Let's consider a two strategy game with a  $2 \times 2$  pay off matrix  $\begin{pmatrix} R & S \\ T & P \end{pmatrix}$  (not necessarily PD)

Let  $u(x,t)$  be density of players using strategy C so that  $v(x,t) = 1 - u(x,t)$  is density of players using strategy D. Now here is how I do the spatial game. Let me look at a weighted nhd around  $x$  + compute the fitness if I played C:

$$f_C(x) = R \int w(y) u(x-y, t) dy + S \int w(y) v(x-y, t) dy$$

and fitness if I played D:

$$f_D(x) = T \int w(y) u(x-y, t) dy + P \int w(y) v(x-y, t) dy$$

Here  $w(y)$  is a symmetric weighting function, say exponential or Gaussian.

I will go from C  $\rightarrow$  D at rate  $\beta$  + from D  $\rightarrow$  C at a rate  $\alpha$  where

$\alpha = H(f_C - f_D)$ ,  $\beta = H(f_D - f_C)$  where  $H$  is some monotonic non-negative function, eg  $1/(1+e^{-\gamma x})$

so, eg if  $f_C > f_D$ , then  $\alpha$  will be bigger than  $\beta$  + there will be a

loss of D + gain of C at  $x$ . :  $\frac{\partial u}{\partial t} = \alpha v - \beta u = \alpha(1-u) - \beta u$

$$\alpha = H(f_C(x) - f_D(x)), \beta = H(f_D(x) - f_C(x))$$

It is now clear how to do this for  $n$  strategy games  
 Let  $A = (a_{ij})$  be payoff matrix. Let  $u_i(x, t)$  be density playing strategy  $i$ . Then

$$f_i(x, t) = \int w(y) \sum_{j=1}^n a_{ij} u_j(x-y, t) dy$$

$$\dot{u}_{ij} = H(f_i - f_j) = \text{rate } j \text{ to } i$$

$$\frac{\partial u_i}{\partial t} = \sum_{j=1}^n a_{ij} u_j \quad \dot{x}_{ij} = H(f_i(x, t) - f_j(x, t))$$

This is not like replicator dynamics, which we will look at later  
 It is a new model. Let's look at equilibria for  $n=2$

$$\frac{du}{dt} = \alpha(1-u) - \beta u = 0 \Rightarrow u = \frac{\alpha}{\alpha + \beta}$$

$$\alpha = H(a_{11}u + a_{12}(1-u) - a_{21}u - a_{22}(1-u))$$

$$= H(a_{12} - a_{22} + (a_{11} - a_{21} + a_{22} - a_{12})u)$$

$$= H(c - (d-c)u)$$

$\beta = H(-c + (d-c)u)$  so as expected all that matters are off diagonal terms, so w.l.o.g we can assume  $A = \begin{bmatrix} 0 & c \\ d & 0 \end{bmatrix}$

For PD  $d > 0, c < 0$ . For bistable game:  $c, d < 0$

so let's try that: Take  $c = -1, d = -4$  for example +  $\gamma = 10$

