

Chapter 3

★ Social evolution is all about ASSORTMENT: who interacts with who.
 What payoff you get depends on who you interact with.

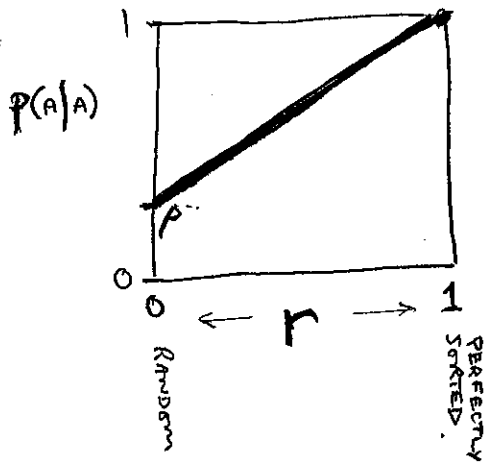
★ Last time we assumed RANDOM MIXING:

if there are 2 types: A, B,
 and p is freq. of A's,
 $\text{prob}(A|A) = p$
 $\text{prob}(B|A) = 1-p$

★ But what if A's...

- tend to seek each other out?
- tend to end up in similar places?

Model with a parameter r : $\begin{cases} r=0 & \text{is RANDOM.} \\ r=1 & \text{is COMPLETELY SORTED} \\ & \text{(like only meets like).} \end{cases}$
 (and $0 < r < 1$ interpolates between these).



$$P(A|A) = r + (1-r)p$$

& Similarly,

$$P(B|B) = r + (1-r)(1-p)$$

& We don't need $P(B|A)$ since it is just $1 - P(A|A)$

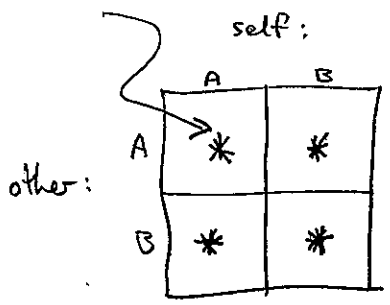
Possible mechanisms for assortment:

- pseudo-physical, like the sorting of stones on a beach by wave action.
- kin recognition: "only cooperate with close relatives..."
- limited dispersal of offspring...?
- other ways ?? ...

IN THIS CASE you could think of r as "relatedness by descent".

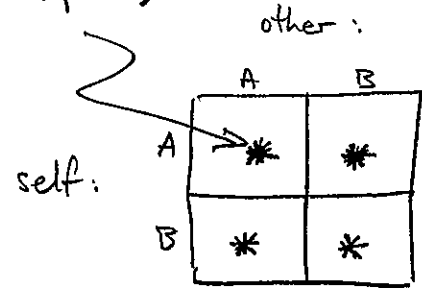
∴ We have probabilities of interaction :

$P(\text{other}^{\text{with}}/\text{self})$



... and payoffs (expected) for each such interaction :

$V(\text{self}^{\text{to}}/\text{other}^{\text{given}})$



So put these together to get expected fitnesses; averaged over interactions

$$W(A) = W_0 + P(A|A) V(A|A) + P(B|A) V(A|B)$$

$$\& W(B) = W_0 + P(B|B) V(B|B) + P(A|B) V(B|A) \quad \checkmark$$

SOME TRIVIAL
TEDIUM

an UGLY (but linear) EQUATION :

$$\begin{aligned} W(A) - W(B) = & [r(1-p) + p][V(A|A) - V(A|B)] + V(A|B) \\ & - [r p + (1-p)][V(B|B) - V(B|A)] - V(B|A) \end{aligned}$$

expected fitness of A's

assortment ("relatedness")

relative frequency of A's

if this r.h.s. > 0, A's will increase at the expense of B's.

Example: Prisoners Dilemma

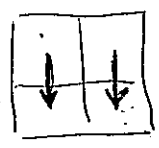
A → "C": cooperater; benefits another by b at cost to self of c
We assume $b > c$ (NB. p is freq. of cooperators.)

B → "D": defector; does nothing...

Payoff matrix = V :

| | | | |
|------|---|-------|------|
| | | other | |
| | | C | D |
| self | C | $b-c$ | $-c$ |
| | D | b | 0 |

Hawk-Dove is the same
Notice structure is



ie. Prefer D regardless of other.

⇒ D-D is rational, but has lower payoff than C-C, hence the "dilemma".

Case 1:

no assortment ($r=0$). Plug that, and the above payoffs into UGLY EQUATION, which becomes

$$W(C) - W(D) = -c \quad (\text{everything else cancels!})$$

This is negative, so D's always fitter than C's, so $p \rightarrow 0$ as expected.

Case 2:

Assume $p \approx 0$ (all Defectors).
At what r value does $W(C)$ become $> W(D)$?
UGLY EQTN, with $p=0$, becomes

$$W(C) - W(D) = rb - c$$

∴ the tipping point is at $r > c/b$

"Hamilton's Rule".

["r" can be relatedness-by-descent, in a kin-selection model. But more generally it's any assortment]

Example: Stag Hunt

(NB. ρ is freq. of stag hunters.)

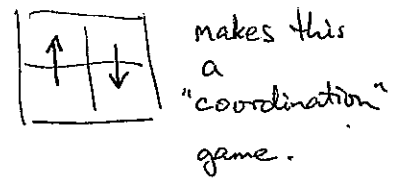
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A → "S": Stag hunter — a pair of stag hunters get benefit s (each) }
 B → "H": Hare hunter — hunts solo, hares have benefit h (to self). } } We assume $s > h$

Payoffs V :

| | | | |
|---|---|-------|---|
| | | other | |
| | | S | H |
| } | S | s | 0 |
| | H | h | h |

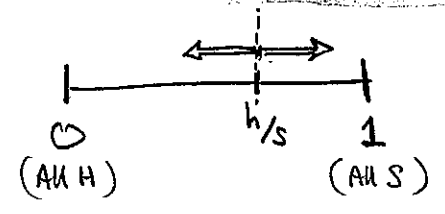
Notice structure is



Case 1 no assortment. ($r = 0$).
 UGLY EQTN becomes....

$$W(S) - W(H) = \rho s - h$$

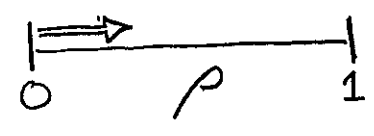
So, Stag hunters increase if their frequency $\rho > h/s$



Case 2 At what r could rare Stag hunters invade?
 $\rho \approx 0$...

$$W(S) - W(H) = r s - h$$

(ie) S can invade a pure H population if $r > h/s$



The Price Equation (simple form)

Consider allele "A", with frequency p .

We showed new generation has:

$$p' = p \frac{W(A)}{\bar{w}}$$

average fitness of the "A" in this population.

average fitness across whole population.

$$\begin{aligned} \text{So } \Delta p &= p' - p \\ &= p \left(\frac{W(A)}{\bar{w}} - 1 \right) \end{aligned}$$

$$\text{or } \bar{w} \Delta p = p W(A) - p \bar{w}$$

we need 3 numbers:
 p , $W(A)$, and \bar{w}

In Terms Of Individuals:

Say $a_i = \begin{cases} 1 & \text{if } i^{\text{th}} \text{ individual carries the A gene.} \\ 0 & \text{otherwise.} \end{cases}$

$$\text{then } p = \frac{1}{n} \sum_{i=1}^n a_i = \bar{a}$$

$$\bar{w} = \frac{1}{n} \sum_{i=1}^n w_i \quad \text{fitness of the } i^{\text{th}} \text{ individual.}$$

$$\text{and } W(A) = \frac{\sum_i w_i a_i}{\sum_i a_i} = \text{av. fitness of those that have } a=1$$

$$\therefore \bar{w} \Delta p = \frac{1}{n} \sum_i w_i a_i - \bar{w} \bar{a}$$

$$= E[w_i a_i] - E[w_i] * E[a_i]$$

$$\bar{w} \Delta p = \text{cov}(w_i, a_i)$$

PRICE'S EQTN, true for any allele.

$$\equiv \text{var}(a_i) * \beta(w_i; a_i)$$

eg. if a_i binary
this is just
 $p(1-p)$

Slope of the regression line relating fitness w_i to gene a_i .



Key points

• sign of $\Delta p = \text{sign of } \beta (w_i; a_i)$

• If our model has a function for fitness in terms of a , we can find β by differentiating the function.
↑ slope of w vs. a

• Example: Prisoner's dilemma.
 $x_i \in \{0, 1\}$ indicator of presence of "cooperator gene".

$$w_i = w_0 + b x_i - c x_i$$

↑ individual that gets to interact with i^{th} .

$$\begin{aligned} \text{Slope: } \beta &= \frac{dw_i}{dx_i} = b \frac{dx_i}{dx_i} - c \\ &= b \left(\frac{\partial x_i}{\partial x_i} \cdot \frac{dx_i}{dx_i} \right) - c \\ &= r b - c \end{aligned}$$

Tricky: method due to Taylor & Franks. See the book....

$\therefore \Delta p$ +ve if

| | |
|-----------|-----------------|
| $r > c/b$ | Hamilton's Rule |
|-----------|-----------------|