

We will discuss the basic building blocks of evolution!

Immortal cells growing w/ perfect nutrients

$$0 \rightarrow \infty \rightarrow 2\infty \rightarrow 4\infty \rightarrow \dots$$

20mins 20m 20m

After t generations (= 20 mins in this example) # cells is 2^t .

Can write a discrete-time formula for this. spiriting/doubling

$$x_{t+1} = 2 \cdot x_t$$

cells at time $t+1$ # cells at time t

$$\Rightarrow x_t = x_0 \cdot 2^t$$

initial # of cells

Can also write continuous time version. $\lambda = \text{rate}$

Suppose cells divide at rate r . (~Exp($\frac{1}{r}$)) \rightarrow splits per time period per cell

$$\dot{x} = \frac{dx}{dt} = r \cdot x$$

cells at time t

$$\Rightarrow x(t) = x_0 e^{rt}$$

If t has units of days, then $r = 1$ split per 20 mins = 3 splits per hour = (3*24) 72 splits per day = 72

After 3 days, one cell will generate $x(3) = e^{3 \cdot 72} = e^{216} \approx 6 \times 10^{92}$ cells!!

Now, lets say cells die at rate d .

$$\dot{x} = rx - dx = (r-d)x$$

$$\Rightarrow x(t) = x_0 e^{(r-d)t}$$

If $r > d$, $x(t) \rightarrow \infty$
 If $r < d$, $x(t) \rightarrow 0$
 If $r = d$, $x(t)$ is constant but unstable

λ is known as reproductive ratio. Need $\lambda > 1$ for population growth

Now, say that there is some upper limit to the # of individuals in a population $K = r \cdot (1/\lambda)$

Logistic Equation: $\dot{x} = \alpha x$ where $\alpha = \beta r$

As x gets larger, β gets smaller, so α gets closer to 0. $\beta = 1 - x/K$ (damping factor)

As x gets smaller, β gets larger, so α gets closer to r .

$$\text{Soln } x(t) = \frac{K x_0 e^{rt}}{K + x_0 (e^{rt} - 1)}$$

Equilibrium of $x(t) \rightarrow K$ as $t \rightarrow \infty$ assuming $x_0 > 0$

Above, we studied reproduction. Now we study selection.

Selection occurs when there are different reproduction rates.

Two types: A & B

$$\dot{x} = a x$$

$$\dot{y} = b y$$

A individuals # B individuals

$$\text{Soln } x(t) = x_0 e^{at}$$

$$y(t) = y_0 e^{bt}$$

Now, suppose population size is held constant. We will use $x(t)$ to be the relative freq of A at time t .

$$x(t) + y(t) = 1 \quad \forall t$$

A proportional amount of A and B are continuously removed from the population. "flux" / "flux" rate

$$\dot{x} = ax - \phi x$$

$$\dot{y} = by - \phi y$$

(*)

What is ϕ ?

We know $x+y=1$, thus $\dot{x}+\dot{y} = \frac{d}{dt}[x+y] = \frac{d}{dt}[1] = 0$

$$\Rightarrow (ax - \phi x) + (by - \phi y) = 0$$

$$\Rightarrow \phi(x+y) = ax + by$$

$$\Rightarrow \phi = ax + by$$

(**)

Also, we know $y=1-x$, thus Eqn (*) becomes:

$$\dot{x} = ax - (ax + by)x$$

$$= ax - (ax + b(1-x))x$$

$$= ax - ax^2 - bx + bx^2$$

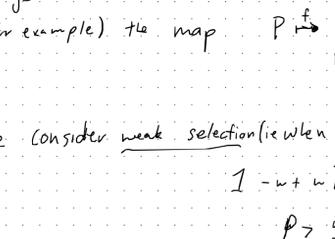
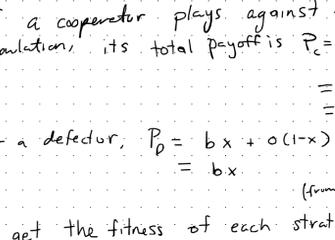
$$= ax - ax^2 - bx + bx^2$$

$$= ax - bx - (ax^2 - bx^2)$$

$$= (a-b)x - (a-b)x^2$$

$$= (a-b)x(1-x)$$

two equilibria $x=0, 1$ but what about stability?



$a=b$ is "degenerate"

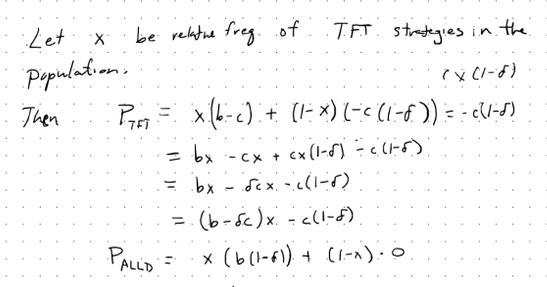
n types (instead of only 2).

$i=1, \dots, n$
 $x_i(t)$ is freq of type i
 population is $\vec{x} = (x_1, \dots, x_n)$ where $\sum_{i=1}^n x_i = 1$
 Let f_i be the fitness (rate of reproduction) of type i .

$$\dot{x}_i = f_i x_i - \phi x_i, \quad i=1, \dots, n$$

(0)

$$\phi = \sum_{i=1}^n x_i f_i = \text{avg fitness of population}$$



n types $\Rightarrow (n-1)$ -dim simplex

Eqn (0) is called the Replicator equation.

Equilibrium where $\dot{x}_i = 0 \Rightarrow (f_i - \phi) x_i = 0 \quad \forall i$
 (non-degenerate) $\Rightarrow (f_i - \sum_{j=1}^n x_j f_j) x_i = 0 \quad \forall i$
 $\Rightarrow ((1-x_i)f_i - \sum_{j \neq i} x_j f_j) x_i = 0$
 $x_i = 1, x_j = 0 \quad \forall j \neq i$

Only stable equilibrium is where $f_i > f_j \quad \forall j \neq i$
 "Survival of the fittest!"

Now we add mutation. (consider two types first, w/ no selection advantages)

u_1 is the rate type A mutates to type B
 u_2 is the rate type B mutates to type A

$$\dot{x} = (1-u_1)x + u_2 y - \phi x$$

$$\dot{y} = (1-u_2)y + u_1 x - \phi y$$

$\phi = 1$ (why?)

$$\text{Thus } x = u_2 - (u_1 + u_2)x$$

stable equilibrium of $x^* = \frac{u_2}{u_1 + u_2}$ when $u_1 + u_2 > 0$

mutation leads to coexistence of A & B

$\sum_{i,j} q_{ij} = 1$
 n types, q_{ij} is mutation rate from i to j

$$\dot{x}_i = \left[\sum_{j=1}^n q_{ji} x_j \right] - \phi x_i, \quad i=1, \dots, n$$

$$Q = [q_{ij}] \Rightarrow \dot{\vec{x}} = \vec{x} Q - \phi \vec{x}$$

Equilibria where $\vec{x}^T Q = \phi \vec{x}^T$

In the case of neutral evolution, $\phi = 1$
 so we are looking for the left hand eigenvector w/ corresponding eigenvalue 1

Now put it all together!

n types $i=1, \dots, n$

frequencies x_i

fitnesses f_i

mutation rates $Q = [q_{ij}]$

$$(00) \quad \dot{x}_i = \left[\sum_{j=1}^n x_j f_j q_{ji} \right] - \phi x_i, \quad i=1, \dots, n$$

where $\phi = \sum_{i=1}^n x_i f_i = \text{avg fitness}$

Eqn. (00) is called the quasispecies equation

- If $f_1 = f_2 = \dots = f_n$, then there is no selection advantage. This is known as "neutral evolution"
- If f_i does not depend on \vec{x} , this is "constant selection"
- If f_i may depend on \vec{x} , this is "frequency-dependent" selection. This is the realm of evolutionary game theory.
- If $Q = I$, there are no mutations.

Donation game

Consider x is relative freq of cooperators $y=1-x$ defectors

If a cooperator plays against everyone in the population, its total payoff is $P_c = [(b-c)x] + [-c(1-x)]$

$$= bx - cx - c + cx = bx - c$$

For a defector, $P_d = bx + 0(1-x) = bx$

To get the fitness of each strategy, we use (for example) the map $P \mapsto 1-w+ wP$ where $w \in [0, 1]$ strength of selection

We consider weak selection (ie when $w \ll 1$)

$$1 - w + wP > 0$$

$$\phi > \frac{w-1}{w} = 1 - \frac{1}{w}$$

Plus up when w is small

$$f_c = 1-w+wP_c$$

$$f_d = 1-w+wP_d$$

Replicator equation gives

$$\dot{x} = (f_c - f_d) x (1-x)$$

$$= \underbrace{[-wc]}_{< 0} x \underbrace{[1-\delta]}_{> 0} (1-x)$$

Direct reciprocity through repeated games

From last lecture, $\delta \in [0, 1]$ is prob. of a next round in a repeated game.

We calculated for the donation game

$$\begin{matrix} & TFT & ALLD \\ TFT & (b-c, -c(1-\delta)) & \\ ALLD & (b(1-\delta), 0) & \end{matrix}$$

Let x be relative freq of TFT strategies in the population.

$$P_{TFT} = x(b-c) + (1-x)(-c(1-\delta)) = -c(1-\delta)$$

$$= bx - cx + cx(1-\delta) - c(1-\delta)$$

$$= bx - \delta cx - c(1-\delta)$$

$$= (b-\delta c)x - c(1-\delta)$$

$$P_{ALLD} = x(b(1-\delta)) + (1-x) \cdot 0 = (1-\delta)bx$$

$$\text{So } f_{TFT} = 1-w+wP_{TFT} \quad \phi = x f_{TFT} + (1-x) f_{ALLD}$$

$$f_{ALLD} = 1-w+wP_{ALLD}$$

$$\dot{x} = (f_{TFT} - f_{ALLD}) x (1-x)$$

$$= \left[(b-\delta c)x - c(1-\delta) - (1-\delta)bx \right] x (1-x)$$

$$= \left[b \cancel{x} - \delta cx - c(1-\delta) - \cancel{bx} + \delta bx \right] x (1-x)$$

$$= \left[\delta(b-c)x - c(1-\delta) \right] x (1-x)$$

γ

$$\gamma = 0 \Leftrightarrow \delta(b-c)x = c(1-\delta)$$

$$\Leftrightarrow \delta(b-c)x + \delta c = c$$

$$\Leftrightarrow \delta[(b-c)x + c] = c$$

$$\Leftrightarrow \delta = \frac{c}{(b-c)x + c} =: \delta^*$$

$$\Rightarrow x^* = \frac{(b-c)}{\delta(b-c)}$$

$x > x^* \Rightarrow \gamma > 0 \Rightarrow$ selection favors TFT $\frac{b-c}{\delta(b-c)}$

$x < x^* \Rightarrow \gamma < 0 \Rightarrow$ selection favors ALLD $\frac{c}{\delta(b-c)}$

$\delta=1 \Rightarrow$

$\delta=0 \Rightarrow$

$\delta=1/2 \Rightarrow$

\Rightarrow

We are missing a crucial component of the dynamics. We have not accounted for drift.

Our deterministic analysis gives us the relative growth rates, but in actuality there is lot of stochasticity involved.

We will introduce a stochastic process for evolution next class!