

Department of Computer Science ETH Zürich

Evolutionary Dynamics

Assignment #01

Lorenzo Gatti, Zahra Karimadini, Aliaksandr Yudzin Thursday 4th October, 2012

1.1 Problem 1: Logistic difference equation

1.1.1 a: Finding the equilibrium points

We can find equilibrium points from

$$f(x,a) = ax(1-x) = x (1.1)$$

as the roots of the quadratic equation:

$$ax^{2} - (a-1)x = x(ax - (a-1)) = 0$$
(1.2)

which are:

$$x_1^* = 0 (1.3)$$

$$x_2^* = \frac{(a-1)}{a} \tag{1.4}$$

The fixed point x_2 is non-negative if $a \ge 1$. One can analyze the local stability of the difference equation 1.1 by examining the partial derivative of f with respect to x evaluated at each fixed point x^* :

$$f' = -ax - a(x - 1) = a - 2ax (1.5)$$

For $|f_x'| < 1$ the equlibrium point is attractive.

$$x_1^* = 0 \rightarrow f'_{(x)} = a - 2a(0) = a \rightarrow |a| < 1$$
 then x_1^* is attractive

$$x_2^* = \frac{(a-1)}{a} \to f'_{(x)} = a - 2a + 2 = -a + 2 \to |2-a| < 1 \text{ for } 1 < a < 3 \text{ then } x_1^* \text{ is attractive}$$

1.1.2 b: Point stability at different values of a

When a = 0.9

$$x_1^* = 0 \rightarrow |a| = |0.9| < 1 \text{ attractive}$$
 (1.6)

$$x_2^* = \frac{(a-1)}{a} \to |2 - 0.9| = |1.1| > 1 \text{ not attractive}$$
 (1.7)

(1.8)

When a = 2.1

$$x_1^* = 0 \to |a| = |2.1| > 1 \text{ not attractive}$$
 (1.9)

$$x_2^* = \frac{(a-1)}{a} \to |2-2.1| = |0.1| < 1 \text{ attractive}$$
 (1.10)

(1.11)

When a = 3..58

$$x_1^* = 0 \rightarrow |a| = |3.58| > 1 \text{ not attractive}$$
 (1.12)

$$x_2^* = \frac{(a-1)}{a} \to |2 - 3.58| = |1.58| > 1 \text{ not attractive}$$
 (1.13)

(1.14)

In a conceptual view we can see for a values in excess of 3.57, the orbits x(t, x0) = x0, x1, x2, ... depend crucially on the initial condition x0. Slight variations in x0 result in dramatically different orbits, an important characteristic of chaos.

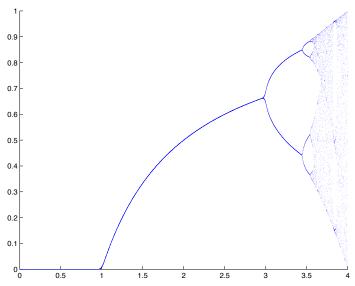


Figure 1.1: Logistic Map Bifurcation Diagram

2.1 Problem 2: Logistic growth in continuous time

2.1.1 a: Finding solution of logistic equation

We have to solve the equation

$$\frac{dx}{dt} = rx(1 - \frac{x}{K}) = r(x - \frac{x^2}{K}) = \frac{rx(K - x)}{K}$$
 (2.15)

using the separation of variables

$$\frac{K}{x(K-x)}dx = rdt ag{2.16}$$

decomposing the right part with partial fractions

$$\frac{K}{x(K-x)} = \frac{A}{x} + \frac{B}{K-x} \tag{2.17}$$

We find A and B

$$A = \frac{K - Bx}{K - x} \tag{2.18}$$

$$B = \frac{K - A(K - x)}{x} \tag{2.19}$$

Supposing A = 1, and according to above, B is also equal to 1, so our partial fraction decomposition is

$$\frac{K}{x(K-x)} = \frac{1}{x} + \frac{1}{K-x} \tag{2.20}$$

Now we have to take the integral from both parts:

$$\int \frac{1}{x} + \frac{1}{K - x} dx = \int r dt \tag{2.21}$$

$$\int \frac{1}{x} + \int \frac{1}{K - x} dx = r \int dt \tag{2.22}$$

$$\ln x - \ln K - x = rt + x_0 \tag{2.23}$$

$$\ln \frac{x}{K - x} = rt + x_0$$
(2.24)

$$\frac{x}{K - x} = x_0 e^{rt} \tag{2.25}$$

Finally the solution

$$x(t) = x_0 K e^{rt} \frac{1}{K + x_0 (e^{rt} - 1)}$$
(2.26)

To prove it we explict the computation behind our results($\lambda = r$):

$$\frac{dx}{dt} = \lambda x (1 - \frac{x}{k}) = \lambda x (\frac{k - x}{k}) \rightarrow \int \frac{k}{x(k - x)} dx = \int \lambda dt$$

$$\int \left[\frac{1}{x} + \frac{1}{k - x} \right] dx = \int \lambda dt \rightarrow \ln|x| - \ln|k - x| = \lambda t + c \rightarrow$$

$$\rightarrow \ln|k - x| - \ln|x| = -\lambda t - x \rightarrow \ln\left|\frac{k - x}{x}\right| = e^{-\lambda t - c} = \frac{k - x}{x} = \pm e^{-\lambda t} e^{-c} \text{ with } A = e^{-c}$$

$$\rightarrow \frac{k - x}{x} = A e^{-\lambda t} = \frac{k}{x} - 1 = A e^{-\lambda t} \rightarrow x = \frac{k}{1 + A e^{-\lambda t}} = \frac{k}{1 + A e^{-\lambda t}} (\frac{e^{\lambda t}}{e^{\lambda t}}) = \frac{k e^{\lambda t}}{e^{\lambda t} + A}$$

$$x_0 = x(0) = \frac{k}{1 + A} = x_0 \rightarrow k = x_0 + A x_0 = k - x_0 = A x_0 \rightarrow A = \frac{k - x_0}{x_0}$$

$$x = \frac{k e^{\lambda t}}{e^{\lambda t} + \frac{k - x_0}{x_0}} = \frac{k x_0 e^{\lambda t}}{e^{\lambda t} x_0 + k - x_0} = \frac{k x_0 e^{\lambda t}}{k + x_0 (e^{\lambda t} - 1)}$$

2.1.2 b: Determining the stability of equilibrium points

To determine the stability of the two equilibria points found solving the equation 2.15 in 0

$$x_1^* = 0 (2.27)$$

$$x_2^* = K (2.28)$$

one has to derive it for *x*:

$$f'(x) = -r(\frac{x}{K} - 1) - \frac{rx}{K} \tag{2.29}$$

Substituting x_1^* and x_2^* into 2.29 yields:

$$f'(x_1^*) = r \text{ if } r < 0 \text{ stable equilibrium}$$
 (2.30)

$$f'(x_2^*) = -r \text{ if } r > 0 \text{ stable equilibrium}$$
 (2.31)

One finds that if $f'(x_1)$ with $x_1 = 0$, then x^* is always stable, while if $forr < 0 \rightarrow f'(x_2) > 0$ with $x_2 = -r x^*$ is repelling.

3.1 Problem 3: Hardy-Weinberg equilibrium

a: frequency of allele B in population

Assuming HWE fot three alleles (allele and genotype frequences):

$$(A + B + 0)^2 = A^2 + B^2 + 0^2 + 2AB + 2A0 + 2B0 = 1$$

We could solve the equation:

$$(A + B + 0)^2 = (p + q + r)^2 = p^2 + q^2 + r^2 + 2pq + 2pr + 2qr = 1$$

And given observable genotype frequences

$$0^2 = \frac{900}{10000} = 0.09 = r^2 \to r = 0.3 \tag{3.32}$$

$$2AB = \frac{2000}{10000} = 0.20 = 2pq \to pq = 0.1 \tag{3.33}$$

$$2AB = \frac{2000}{10000} = 0.20 = 2pq \to pq = 0.1$$

$$A^2 + 2A0 = \frac{1600}{10000} = 0.16 = p^2 + 2pr$$
(3.33)

(3.35)

$$p^2 + 2 * 0.3 * p = 0.16 (3.36)$$

$$p_1 = \frac{1}{5} \tag{3.37}$$

$$p_2 = -\frac{4}{5}$$
 not acceptable (3.38)

(3.39)

$$q = \frac{0.1}{0.2} = 0.5 \tag{3.40}$$

So, the frequency for *B* allele is equal to 0.5

3.1.2 b: HW Equilibrium

The equilibrium is given by $p^* = P^2 + pq + pr$

$$A^2 = \frac{1500}{10000} = 0.15 = p^2 \to p = 0.39 \tag{3.41}$$

$$p^2 + 2pr = \frac{1600}{10000} = 0.16 \to 0.15 + 2pr = 0.16 \to pr = \frac{0.01}{2} = 0.005$$
 (3.42)

$$2pq = \frac{2000}{10000} = 0.2 \to pq = 0.1 \tag{3.43}$$

(3.44)

$$p^* = 0.15 + 0.0005 + 0.1 = 0.255$$

So, for the next round of random mating, the number of genotype $AA = (P^*)^2 * 10000 = (0.255)^2 * 10000 = 650 \neq 1500$. The first generation is not in Hardy-Weinberg equilibrium.

Hence, assuming population carrying AA is 1500, we are not at the equilibrium of HW ($p_b^2 = 0.15 \neq p_a^2 = 0.04$)

4.1 Problem 4: Sequence alphabets

4.1.1 A: Amino acid sequences

According to the number of amino acids c, $c_1 = 20$ standard aminoacids $|c_2| = 22$ standard + non-canonical we consider there are:

for
$$c_1 = 20$$
 then $s_{space} = c_{1,\ell}^L 20^{50} = 1.13 * 10^{65}$ (4.45)

for
$$c_2 = 22$$
 then $s_{space} = c_2^L, 22^{50} = 1.32 * 10^{67}$ (4.46)

Since each aminoacid is not codified by an unique codon (3 nt), the number of unique DNA sequences is much larger then unique aminoacids sequences.

4.1.2 B: Amino acids sequences and DNA sequences

Each amino acid is codified by 3 *nt*, so for a sequence of 50 amino acids we need 150 *nt* or 153 *nt* if we consider the codon stop at the end of the codifying sequence.

for
$$L = 150$$
 then $d_{space} = c^L = 4^{150} = 2.04 * 10^{90}$ (4.47)

for
$$L = 153$$
 then $d_{space} = c^L = 4^{153} = 1.30 * 10^{92}$ (4.48)

Comparison between the number of unique amino acid sequence and DNA unique sequences:

$$\frac{4^{150}}{20^{50}} = 1.8 * 10^{25} = (\frac{64}{20})^{50}$$
 more DNA sequences

5.1 Problem 5: Random sequences

5.1.1 a: Average and Expected distance

If the Hamming distance is the number of coordinates where two vectors $x = (x_1, ..., x_n)$ and $y = (y_1, ..., y_n)$ of length N differ

$$d_H(x,y) = \sum_{i=1}^{n} |x_i - y_i|$$

For a set $A \subseteq F_2^n$, |A| denotes the cardinality of A. The average distance in A is defined by

$$dist(A) = \frac{1}{|A|^2} \sum_{x \in A} \sum_{y \in A} d_H(x, y)$$
 (5.49)

Hence

$$dist(F_2^n) = \frac{1}{2^{2n}} \sum_{x \in F_2^n} \sum_{y \in F_2^n} d_H(x, y) = \frac{n2^{n-1}2^n}{2^{2n}} = \frac{n}{2}$$
 (5.50)

Let *V* be some finite set with *q* elements where $n \ge 1$ and V^n is the dimension of sequence space. *P* is the common probability distribution. Then, we have

$$\frac{n(q-1)}{q} - L(P) \le Ed_H(X,Y) \le \frac{n(q-1)}{q}$$

where L(P) measures how skewly P is distributed as

$$L(P) = q^{n-1} \sum_{x \in V^n} \left[P(x) - \frac{1}{q^n} \right]^2$$

If $2 \le q \ge 4$, (in DNA q = 4) then

$$Dd_H(X,Y) \le \frac{n(q-1)}{q^2} + \frac{2}{q}L(P)$$

Hence

$$\frac{n(q-1)}{q^2} \le Ed_H(X,Y) \le \frac{n(q-1)}{q^2} + \frac{2}{q}L(P)$$

This gives out that for two i.i.d. random sequences with the common probability distribution P, we have for n = L and q = 4

$$Ed_H(X,Y) = \frac{n(q-1)}{q} = \frac{L(4-1)}{4} = \frac{3}{4}L$$
 (5.51)

So, for binary sequences we can compute the average distance with equation 5.51 as well.

5.1.2 b: Number of sequences at a certain distance

With given sequences of length *L* for Hamming distance *K* and alphabet *A*:

$$c(L,K) = (A-1)^K \frac{L!}{(L-K)!K!}$$

For a binary sequence (A = 2) of length L at distance K = 1:

$$c(L, K) = (2-1)^K \frac{L!}{(L-K)!K!}$$
 for $K \le L$

$$c(L,K) = (2-1)^{1} \frac{L!}{(L-1)!1!} = \frac{L!}{(L-1)!} = L$$

For DNA sequences (A = 4) of length L at distance K = 1:

$$c(L,K) = (4-1)^{K} \frac{L!}{(L-K)!K!}$$

$$c(L,K) = 3^{K} \frac{L!}{(L-K)!K!} \text{ for } K \le L$$

$$c(L,K) = 3^{1} \frac{L!}{(L-1)!1!} = 3L$$

5.1.3 c: Replication of sequences

Considering H_{ij} as d and $q_i j$ as q:

$$q_{ij} = p^{d} (1 - p)^{L - d}$$

$$\ln q = \ln(p^{d} (1 - p)^{L - d}) = \ln p^{d} + \ln(1 - p)^{L - d}$$

$$d \ln p + (l - d) \ln(1 - p) = d \ln p - l \ln(1 - p) - d \ln(1 - p)$$

$$d(\ln p - \ln(1 - p)) = \ln q - l \ln(1 - p)$$

$$d = \frac{\ln q - l \ln(1 - p)}{\ln p - \ln(1 - p)}$$

6.1 Problem 6: Quasispecies

6.1.1 a: Find mutation-selection matrix w

$$\frac{dx}{dt} = wx - \Phi x$$

If Q is error free then Q = I but here Q is not only error free but also the genotypes are replicated with probability q. So,

$$Q = (q_{ij}) = (f_j q_{ji}) = \begin{bmatrix} q & 1 - q \\ 1 - q & q \end{bmatrix}$$

with
$$f_1 = 1$$
 and $f_0 > 1 \rightarrow W = (w_{ij}) = \begin{pmatrix} w_{0i} & w_{01} \\ w_{10} & w_{11} \end{pmatrix} = \begin{pmatrix} f_0 q_{00} & f_1 q_{01} \\ f_0 q_{10} & f_1 q_{11} \end{pmatrix} = \begin{pmatrix} f_0 q & 1 - q \\ f_0 (1 - q) & q \end{pmatrix}$

 Φ is the largest eigen value of w, while x is the eigen vector of w.

To find eigen values and eigen vector: $|w - \lambda I| = 0$

$$\begin{bmatrix} f_0q - \lambda & 1 - q \\ f_0 - f_0q & q - \lambda \end{bmatrix} = 0 \to (f_0q - \lambda)(q - \lambda) - (f_0 - f_0q)(1 - q) \to$$

$$f_0q^2 - f_0q\lambda - \lambda q + \lambda^2 - f_0 + f_0q + f_0q - f_0q^2 = 0$$

$$\lambda^2 - (f_0q + q)\lambda + (2f_0q - f_0) = 0$$

$$\begin{cases} \lambda_1 = \frac{(f_0q + q) + \sqrt{(f_0q + q)^2 - 4(2f_0q - f_0)}}{2} \\ \lambda_2 = \frac{(f_0q + q) - \sqrt{(f_0q + q)^2 - 4(2f_0q - f_0)}}{2} \end{cases}$$

6.1.2 b: Finding nontrivial solutions

 Φ is the largest eigen value of, so $\Phi = \lambda_1$ **nontrivial solution**.

6.1.3 c: Equilibrium point for $f_0 = f_1 = 1$

if $f_0 = f_1 = 1$ then

$$\Phi = \frac{2q + \sqrt{(2q)^2 - q(2q - 1)}}{2} = \frac{2q + \sqrt{4q^2 - 4(2q - 1)}}{2} = \frac{2q + 2\sqrt{q^2 - 2q + 1}}{2} = q + \sqrt{(q - 1)^2} = q + (q - 1)$$

$$\begin{cases} \Phi_1 = 2q - 1 \\ \Phi_2 = 1 \end{cases}$$

$$\begin{pmatrix} x_0' \\ x_1' \end{pmatrix} = \begin{pmatrix} q & 1-q \\ 1-q & q \end{pmatrix} \begin{pmatrix} x_0 \\ x_1 \end{pmatrix} - \Phi \begin{pmatrix} x_0 \\ x_1 \end{pmatrix}$$

if $\Phi = 1$, then:

$$\begin{pmatrix} x_0' \\ x_1' \end{pmatrix} = \begin{pmatrix} qx_0 + (1-q)x_1 \\ (1-q)x_0 + qx_1 \end{pmatrix} - \begin{pmatrix} x_0 \\ x_1 \end{pmatrix}$$

$$x'_0 = qx_0 + (1 - q)x_1 - x_0 = f(x_0, x_1)$$

$$x'_1 = (1 - q)x_0 + qx_1 - x_1 = g(x_0, x_1)$$

To find the equilibrium points $f(x_0, x_1) = g(x_0, x_1) = 0$. So

$$qx_0^* + x_1^* - qx_1^* - x_0^* = 0 \to (1 - q)x_0^* = (1 - q)x_1^*$$

$$1 - q)x_0^* + qx_1^* - x_1^* = 0 \to (1 - q)x_0^* = (1 - q)x_1^*$$

$$\to x_0^* = \frac{1 - q}{1 - q}x_1^* = x_1^* \to x_0^* = x_1^*$$

Population type 0 is equal to population type $1 = \frac{1}{2}$

6.1.4 d

 $f_0 \ge 0$ and q = 1

$$q = 1 \rightarrow \Phi = \frac{(f_0 + 1) + \sqrt{(f_9 + 1)^2 - 4(2f_0 - f_0)}}{2} = \frac{(f_0 + 1) + \sqrt{(f_0 + 1)^2 - 4f_0}}{2} = \frac{(f_0 + 1) + (f_0 - 1)}{2}$$

$$\begin{cases} \Phi_1 = \frac{f_0 + 1 + f_0 - 1}{2} = f_0 \\ \Phi_2 = \frac{f_0 + 1 - f_0 + 1}{2} = 1 \end{cases}$$

Since Φ should be the largest and $f_0 \gg 1$ we choose Φ_1 .

$$\begin{pmatrix} x_0' \\ x_1' \end{pmatrix} = \begin{pmatrix} q & 1-q \\ 1-q & q \end{pmatrix} \begin{pmatrix} x_0 \\ x_1 \end{pmatrix} - f_0 \begin{pmatrix} x_0 \\ x_1 \end{pmatrix} = \begin{pmatrix} q & 1-q \\ 1-q & q \end{pmatrix} \begin{pmatrix} x_0 \\ x_1 \end{pmatrix} - \begin{pmatrix} f_0 x_0 \\ f_0 x_1 \end{pmatrix}$$

$$x_0' = qx_0 + (1-q)x_1 - f_0x_0 = f(x_0, x_1)$$

$$x_1' = (1-q)x_0 + qx_1 - f_0x_1 = g(x_0, x_1)$$

To find the equilibrium point $f(x_0, x_1) = g(x_0, x_1) = 0$

$$qx_0^* + x_1^* - qx_1^* - f_0x_0^* = 0 \to (f_0 - q)x_0^* = (1 - q)x_1^*$$

$$1 - q)x_0^* + qx_1^* - f_0x_1^* = 0 \to (f_0 - q)x_0^* = (f_0 - q)x_1^*$$

$$\to x_0^* = \frac{f_0 - q}{1 - q} = \frac{f_0 - 1}{0} \to x_0^* = \infty \text{ whole the population is type 1}$$

$$\to x_1^* = \frac{1 - q}{f_0 - q} = \frac{1 - 1}{f_0 - 1} \to x_1^* = 0 \text{ no type 0 in the population}$$