

Stability of equilibrium points

Higher-Order Scalar Difference Equations

Def. A point x^* is called an equilibrium point for the difference equation

$$x_{n+k} = f(x_{n+k-1}, \dots, x_n)$$

$$\text{if } x^* = f(x^*, \dots, x^*)$$

Linear case

Consider the k th-order difference equation

$$x_{n+k} + p_1 x_{n+k-1} + p_2 x_{n+k-2} + \dots + p_k x_n = 0$$

$$x^* = 0$$

characteristic equation

$$q^k + p_1 q^{(k-1)} + p_2 q^{(k-2)} + \dots + p_k = 0$$

Theorem. $x^* = 0$ is asymptotically stable if and only if $|q| < 1$ for all roots of characteristic equation

Nonlinear case

$$x_{n+k} = f(x_{n+k-1}, \dots, x_n)$$

$$p_i = \frac{\partial}{\partial u_i} f(x^*, \dots, x^*)$$

linearized equation

$$y_{n+k} = p_1 y_{n+k-1} + p_2 y_{n+k-2} + \dots + p_k y_n$$

characteristic equation corresponding to the linearized equation is

$$q^k = p_1 q^{(k-1)} + p_2 q^{(k-2)} + \dots + p_k$$

Theorem (The Linearized Stability Result).

Suppose that f is continuously differentiable on an open neighborhood G from $R^{(k+1)}$ of (x^*, x^*, \dots, x^*) , where x^* is an equilibrium point of the nonlinear difference equation Then the following statements are true:

- If all the characteristic roots of the linearized characteristic equation lie inside the unit disk in the complex plane, then the equilibrium point x^* is (locally) asymptotically stable.
- If at least one characteristic root of the linearized characteristic equation is outside the unit disk in the complex plane, the equilibrium point x^* is unstable.
- If one characteristic root is on the unit disk and all the other characteristic roots are either inside or on the unit disk, then the equilibrium point x^* may be stable, unstable, or asymptotically stable.

Example

The second order Beverton.Holt model

$$x_{n+1} = \frac{r K (\alpha x_n + \beta x_{n-1})}{K + (r - 1) x_{n-1}}$$

has been used to model populations of bottom-feeding fish, where $\alpha + \beta = 1$, $r > 0$, $K > 0$. These species have very high fertility rates and very low survivorship to adulthood. Furthermore, recruitment is

essentially unaffected by fishing. In this model, the future generation x_{n+1} depends not only on the present generation x_n but also on the previous generation x_{n-1} .

```
> f:=(u,v)->r*K*(alpha*u+beta*v)/(K+(r-1)*v);
f: (u, v) →  $\frac{r K (\alpha u + \beta v)}{K + (r - 1) v}$ 

> eq:=x=f(x,x);
eq := x =  $\frac{r K (\alpha x + \beta x)}{K + (r - 1) x}$ 

> eqp:=solve(eq,x);
eqp := 0,  $\frac{K (-1 + r \alpha + r \beta)}{r - 1}$ 

> x1:=eqp[1];x2:=simplify(subs(beta=1-alpha,eqp[2]));
x1 := 0
x2 := K

> p1:=D[1](f)(x1,x1);
p1 := r α

> p2:=D[2](f)(x1,x1);
p2 := r β

> lineq:=y(n+1)=p1*y(n)+p2*y(n-1);
lineq := y(n + 1) = r α y(n) + r β y(n - 1)

> chareq:=q^2=p1*q+p2;
chareq := q2 = r α q + r β

> solve(chareq,q);
 $\frac{r \alpha}{2} + \frac{\sqrt{r^2 \alpha^2 + 4 r \beta}}{2}, \frac{r \alpha}{2} - \frac{\sqrt{r^2 \alpha^2 + 4 r \beta}}{2}$ 
```

Analysing the roots $x_1^*=0$ is locally asymptotically stable if $0 < r < 1$.

```
> p1:=D[1](f)(x2,x2);

$$p1 := \frac{r K \alpha}{K + K(r - 1)}$$


> p1:=simplify(p1);

$$p1 := \alpha$$


> p2:=D[2](f)(x2,x2);

$$p2 := \frac{r K \beta}{K + K(r - 1)} - \frac{r K (\alpha K + \beta K) (r - 1)}{(K + K(r - 1))^2}$$


> p2:=simplify(p2);

$$p2 := \frac{-r \alpha + \alpha + \beta}{r}$$


> p2:=subs(beta=1-alpha,p2);

$$p2 := \frac{-r \alpha + 1}{r}$$


> lineq:=y(n+1)=p1*y(n)+p2*y(n-1);

$$\text{lineq} := y(n + 1) = \alpha y(n) + \frac{(-r \alpha + 1) y(n - 1)}{r}$$


> chareq:=q^2=p1*q+p2;

$$\text{chareq} := q^2 = q \alpha + \frac{-r \alpha + 1}{r}$$


> solve(chareq,q);

$$\frac{r \alpha + \sqrt{r^2 \alpha^2 - 4 r^2 \alpha + 4 r}}{2 r}, \frac{r \alpha - \sqrt{r^2 \alpha^2 - 4 r^2 \alpha + 4 r}}{2 r}$$

```

Analysing the roots we get that the equilibrium $x_2^* = K$ is locally asymptotically stable if and only if $r > 1$.

Remark. If $r > 1$ then $x_2^* = K$ is globally asymptotically stable

Numerical simulations

```
> K:=100;r:=0.8;alpha:=0.5;beta:=1-alpha;N:=100;

$$K := 100$$


$$r := 0.8$$


$$\alpha := 0.5$$


$$\beta := 0.5$$


$$N := 100$$

```

```

> f(u,v);

$$\frac{80.0 (0.5 u + 0.5 v)}{100 - 0.2 v}$$


> x[0]:=20;x[1]:=40;

$$x_0 := 20$$


$$x_1 := 40$$


> for i from 0 to N-2 do
  x[i+2]:=f(x[i+1],x[i])
end do:
> plot([[n,x[n]]$n=0..N],style=point,symbol=circle);

> K:=100;r:=1.2;alpha:=0.5;beta:=1-alpha;N:=100;

$$K := 100$$


$$r := 1.2$$


$$\alpha := 0.5$$


$$\beta := 0.5$$


$$N := 100$$


> f(u,v);

$$\frac{120.0 (0.5 u + 0.5 v)}{100 + 0.2 v}$$

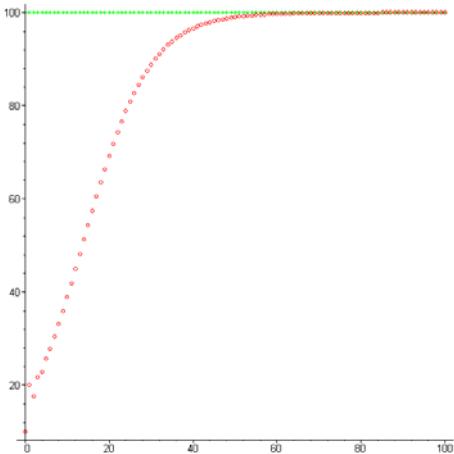

> x[0]:=10;x[1]:=20;

$$x_0 := 10$$


$$x_1 := 20$$


> for i from 0 to N-2 do
  x[i+2]:=f(x[i+1],x[i])
end do:
>
```

```
plot([[n,x[n]]$n=0..N],[[n,K]$n=0..N]],style=[point,point],symbol=[circle,cross],color=[red,green]);
```



```
> restart;
>
```

Autonomous Systems of Difference equations

Linear case

$$x_{n+1} = A x_n$$

Theorem. The following statements hold:

- (i) The zero solution is stable if and only if $|\lambda| \leq 1$ and the eigenvalues of unit modulus are simple
- (ii) The zero solution is asymptotically stable if and only if $|\lambda| < 1$ for all eigenvalues of A.

Nonlinear case

$$x_{n+1} = f(x_n)$$

$x^* = f(x^*)$ equilibrium point

linearized system

$$y_{n+1} = \text{Jacobian}(f)(x^*) y_n$$

Theorem. The following statements hold:

- (i) The x^* is locally asymptotically stable if and only if $|\lambda| < 1$ for all eigenvalues of $\text{Jacobian}(f)(x^*)$.
- (ii) The x^* is unstable if there exists an eigenvalue of $\text{Jacobian}(f)(x^*)$ such that $|\lambda| > 1$.

Examples

Pielou logistic delay equation

$$x_{n+1} = \frac{\alpha x_n}{1 + \beta x_{n-1}}$$

To change this difference equation to a planar system, we let

$$x1(n) = x_{n-1}, \quad x2(n) = x_n$$

$$\begin{bmatrix} x1(n+1) \\ x2(n+1) \end{bmatrix} = \begin{bmatrix} x2(n) \\ \frac{\alpha x2(n)}{1 + \beta x1(n)} \end{bmatrix}$$

```
> with(linalg):
```

```
Warning, the protected names norm and trace have been redefined and unprotected
```

```
> f:=(u,v)->vector([v,alpha*v/(1+beta*u)]);
```

$$f := (u, v) \rightarrow \begin{bmatrix} v, \frac{\alpha v}{1 + \beta u} \end{bmatrix}$$

```
> f(u,v)[1];f(u,v)[2];
```

$$\begin{aligned} v \\ \frac{\alpha v}{1 + \beta u} \end{aligned}$$

```
> eqp:=solve({f(u,v)[1]=u,f(u,v)[2]=v},{u,v});
```

$$eqp := \{ v = 0, u = 0 \}, \{ v = \frac{\alpha - 1}{\beta}, u = \frac{\alpha - 1}{\beta} \}$$

```
> jacobian(f(u,v),[u,v]);
```

$$\begin{bmatrix} 0 & 1 \\ -\frac{\alpha v \beta}{(1 + \beta u)^2} & \frac{\alpha}{1 + \beta u} \end{bmatrix}$$

```
> A1:=subs(u=0,v=0,jacobian(f(u,v),[u,v]));
```

$$A1 := \begin{bmatrix} 0 & 1 \\ 0 & \alpha \end{bmatrix}$$

```
> eigenvals(A1);
```

$$0, \alpha$$

```
> A2:=subs(u=(alpha-1)/beta,v=(alpha-1)/beta,jacobian(f(u,v),[u,v]));
```

$$A2 := \begin{bmatrix} 0 & 1 \\ -\frac{\alpha-1}{\alpha} & 1 \end{bmatrix}$$

```
> eigenvals(A2);

$$\frac{\alpha + \sqrt{-3\alpha^2 + 4\alpha}}{2\alpha}, \frac{\alpha - \sqrt{-3\alpha^2 + 4\alpha}}{2\alpha}$$

```

Numerical simulation

```
> x1[0]:=10;x2[0]:=20;
```

$$x1_0 := 10$$

$$x2_0 := 20$$

```
> alpha:=0.8;beta:=1;N:=100;
```

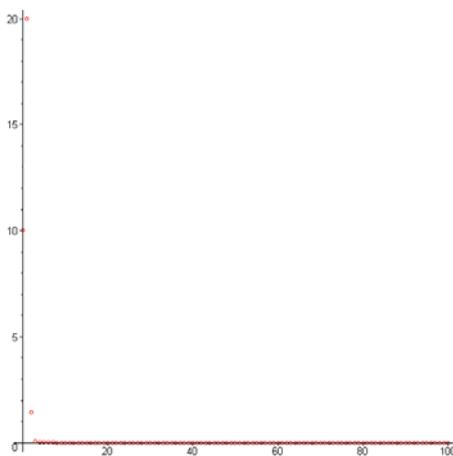
$$\alpha := 0.8$$

$$\beta := 1$$

$$N := 100$$

```
> for i from 0 to N-1 do
  x1[i+1]:=f(x1[i],x2[i])[1];
  x2[i+1]:=f(x1[i],x2[i])[2];
end do:
```

```
> plot([[n,x1[n]]$n=0..N],style=point,symbol=circle);
```



```
> plot([[n,x2[n]]$n=0..N],style=point,symbol=circle);
```

```

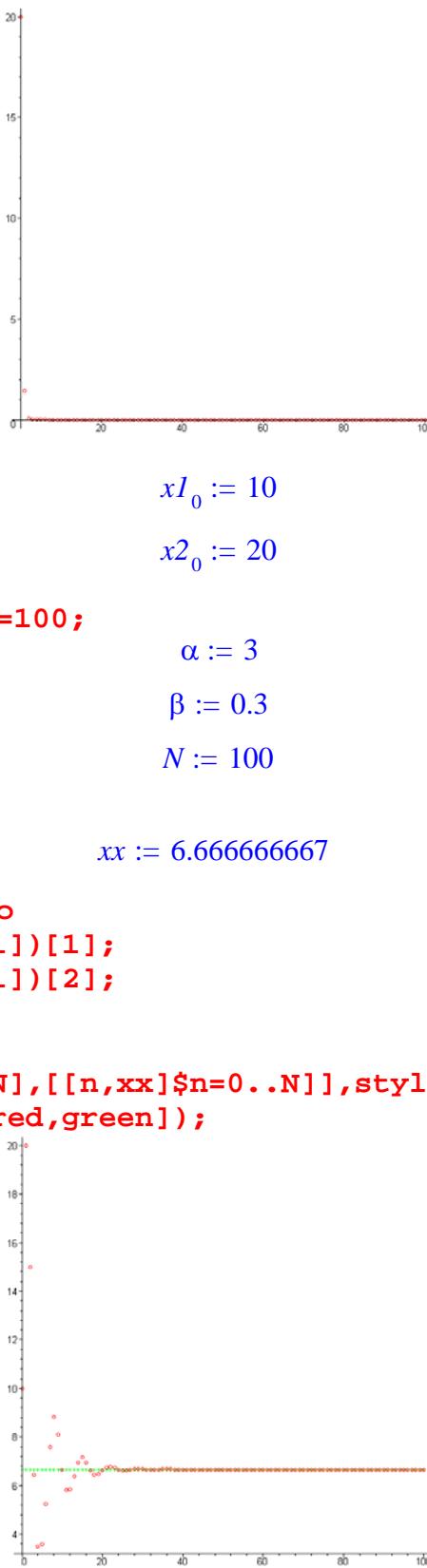
> x1[0]:=10;x2[0]:=20;
           $x1_0 := 10$ 
           $x2_0 := 20$ 

> alpha:=3;beta:=0.3;N:=100;
           $\alpha := 3$ 
           $\beta := 0.3$ 
           $N := 100$ 

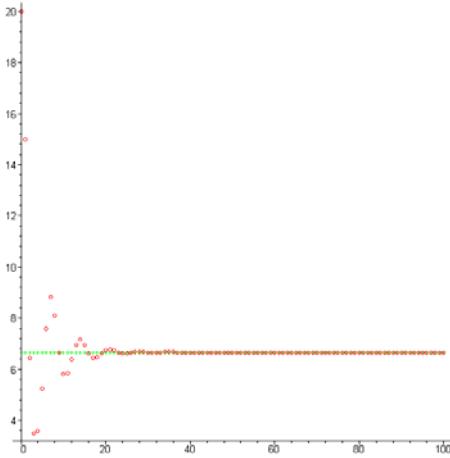
> xx:=(alpha-1)/beta;
           $xx := 6.666666667$ 

> for i from 0 to N-1 do
    x1[i+1]:=f(x1[i],x2[i])[1];
    x2[i+1]:=f(x1[i],x2[i])[2];
end do:
>
plot([[n,x1[n]]$n=0..N],[[n,xx]$n=0..N]],style=[point,point],symbol=
[circle,cross],color=[red,green]);

```



```
> plot([[n,x2[n]]$n=0..N],[[n,xx]$n=0..N]],style=[point,point],  
symbol=[circle,cross],color=[red,green]);
```



```
>
```