

```

> #####
## DMB.txt Save this file as DMB.txt to use it,          #
# stay in the                                           #
## same directory, get into Maple (by typing: maple <Enter> ) #
## and then type: read `DMB.txt` <Enter>                #
## Then follow the instructions given there            #
##                                                    #
## Written by Doron Zeilberger, Rutgers University ,    #
## DoronZeil at gmail dot com                          #
#####

```

```

print( `First Written: Nov. 2021 ` ) :
print( ) :
  print( `This is DMB.txt, A Maple package to explore Dynamical models in Biology (both
  discrete and continuous)` ) :
  print( `accompanying the class Dynamical Models in Biology, Rutgers University. Taught by
  Dr. Z. (Doron Zeilberger) ` ) :

```

```

print( ) :
print( `The most current version is available on WWW at:` ) :
print( ` http://sites.math.rutgers.edu/~zeilberg/tokhniot/DMB.txt . ` ) :
print( `Please report all bugs to: DoronZeil at gmail dot com . ` ) :
print( ) :
print( `For general help, and a list of the MAIN functions, ` ) :
print( ` type "Help():". For specific help type "Help(procedure_name);" ` ) :
print( ` ` ) :

```

```

print( `-----` ) :
print( `For a list of the supporting functions type: Help1();` ) :
print( `For help with any of them type: Help(ProcedureName);` ) :
print( ) :
print( `-----` ) :

```

```

  print( `For a list of the functions that give examples of Discrete-time dynamical systems (some
  famous), type: HelpDDM();` ) :
print( `For help with any of them type: Help(ProcedureName);` ) :
print( ) :
print( `-----` ) :

```

```

  print( `For a list of the functions continuous-time dynamical systems (some famous) type:
  HelpCDM();` ) :
print( `For help with any of them type: Help(ProcedureName);` ) :
print( ) :
print( `-----` ) :

```

*with(LinearAlgebra) :*

*Help1 :=proc( )*

**if** *args = NULL* **then**

*print( `The SUPPORTING procedures are` ) :*

*print( `IsContStable, IsDisStable, JAC, PhaseDiag, RandNice, TimeSeriesE, ToSys` ) :*

**else**

*Help(args) :*

**fi:**

**end:**

*HelpDDM :=proc( )*

**if** *args = NULL* **then**

*print( `The procedures giving discrete-time dynamical systems (some famous), by giving the underlying transformations, followed by the list of variables used are:` ) :*

*print( `AllenSIR, AllenSIRg, Hassell, HW, HWg, May75, NicholsonBailey, NicholsonBaileyM, RT, Valery` ) :*

**else**

*Help(args) :*

**fi:**

**end:**

*HelpCDM :=proc( )*

**if** *args = NULL* **then**

*print( `The procedures giving the underlying transformations, followed by the list of variables used are:` ) :*

*print( `ChemoStat, GeneNet, Lotka, RandNice, SIRS , SIRSdemo, Volterra, VolterraM ` ) :*

**else**

*Help(args) :*

**fi:**

**end:**

```
Help :=proc( )
if args = NULL then
```

```
print( `DMB.txt: A Maple package for exploring Dynamical models in Biology ` ) :
```

```
print( `The MAIN procedures are ` ) :
```

```
print( ` ComK, Dis, EquP, FP, Orb, OrbF, Orbk, OrbkF, PhaseDiag, SEquP, SFP,
TimeSeries ` ) :
```

```
elif nargs = 1 and args[1] = AllenSIR then
```

```
print( `AllenSIR(a,b,c,x,y): The Linda Allen discrete SIR model given in https://sites.math.rutgers.edu/~zeilberg/Bio21/AllenSIR.pdf` ) :
```

```
print( `with parameters a,b,c. try: ` ) :
```

```
print( `AllenSIR(1,1/3,1/3,x,y); ` ) :
```

```
print( `WARNING: To get the long-term behavior, use OrbF NOT Orb (or else Maple will go
for ever) ` ) :
```

```
print( `Try the following: ` ) :
```

```
print( `F:=AllenSIR(1,0.3,0.3,x,y);a:=OrbF(F,[x,y],[1.0, 2.0],1000,1010)[-1];evalf(subs({x=
a[1],y=a[2]},F)-a); ` ) :
```

```
elif nargs = 1 and args[1] = AllenSIRg then
```

```
print( `AllenSIRg(a,b,c,alpha,beta,x,y): The GENERALIZED Linda Allen discrete SIR model
given in https://sites.math.rutgers.edu/~zeilberg/Bio21/AllenSIR.pdf` ) :
```

```
print( `with parameters a,b,c. Try: ` ) :
```

```
print( `where the exponents of x_n and y_n are alpha and beta. Note that ` ) :
```

```
print( `AllenSIRg(a,b,c,1,1,x,y) is the same as AllenSIR(a,b,c,x,y): Try: ` ) :
```

```
print( `AllenSIRg(1,1/3,1/3,1.2,1.2,x,y); ` ) :
```

```
elif nargs = 1 and args[1] = ChemoStat then
```

```
print( `ChemoStat(N,C,a1,a2): The Chemostat continuous-time dynamical system with N=
Bacterial population density, and C=nutrient Concentration in growth chamber (see Table
4.1 of Edelstein-Keshet, p. 122) ` ) :
```

```
print( `with paramerts a1, a2, Equations (19a, (19b) in Edelestein-Keshet p. 127 (section
4.5, where they are called alpha1, alpha2). a1 and a2 can be symbolic or numeric. Try: ` ) :
```

```
print( `ChemoStat(N,C,a1,a2); ` ) :
```

```
print( `ChemoStat(N,C,2,3); ` ) :
```

```
elif nargs = 1 and args[1] = ComK then
```

```
print( `ComK(F,x,K): inputs a transformation F in the list of variables x, outputs the
composition of F with itself K times. Try: ` ) :
```

```
print( `ComK([k*x*(1-x)],[x],2); ` ) :
```

```
print( `ComK([x*(1-y),y*(1-x)],[x,y],4); ` ) :
```

```
elif nargs = 1 and args[1] = Dis then
```

```

print( `Dis(F,x,pt,h,A): Inputs a transformation F in the list of variables x` ) :
  print( `The approximate orbit of the Dynamical system approximating the the autonomous
  continuous dynamical process ` ) :
print( `dx/dt=F[1](x(t)) by a discrete time dynamical system with step-size h from t=0 to t=A` ) :
print( `Try: ` ) :
print( `Dis([x*(1-y),y*(1-x)],[x,y],[0.5,0.5], 0.01, 10); ` ) :

```

**elif nargs = 1 and args[1] = EquP then**

```

  print( `EquP(F,x): Given a transformation F in the list of variables finds all the Equilibrium
  points of the continuous-time dynamical system  $x'(t)=F(x(t))$  ` ) :
print( `EquP([5/2*x*(1-x)],[x]); ` ) :
print( `EquP([y*(1-x-y),x*(3-2*x-y)],[x,y]); ` ) :

```

**elif nargs = 1 and args[1] = FP then**

```

  print( `FP(F,x): Given a transformation F in the list of variables finds all the fixed point of
  the transformation  $x \rightarrow F(x)$ , i.e. the set of solutions of ` ) :
print( `the system {x[1]=F[1], ..., x[k]=F[k]}. Try: ` ) :
print( `FP([5/2*x*(1-x)],[x]); ` ) :
print( `evalf(FP([(1+x+y)/(2+3*x+y), (3+x+2*y)/(5+x+3*y)],[x,y])); ` ) :

```

**elif nargs = 1 and args[1] = GeneNet then**

```

  print( `GeneNet(a0,a,b,n,m1,m2,m3,p1,p2,p3): The contiuous-time dynamical system, with
  quantities m1,m2,m3,p1,p2,p3, due to M. Elowitz and S. Leibler ` ) :
print( `described in the Ellner-Guckenheimer book, Eq. (4.1) (chapter 4, p. 112) ` ) :
  print( `and paramereers a0 (called alpha_0 there),a (called alpha there), b (called beta there)
  and n. Try: ` ) :
print( `GeneNet(0,0.5,0.2,2,m1,m2,m3,p1,p2,p3); ` ) :

```

**elif nargs = 1 and args[1] = Hassell then**

```

  print( `Hassell(L,a,b,N): The discrete-time, single-species dynamical model of Hassell (1975)
  given by Eq. (13) in Edelstein-Keshet section 3.1 (p. 75) ` ) :
  print( `where the variable is N (the population), and the parameters are L (called Lambda
  there), a, and b ` ) :
print( `Try: ` ) :
print( `Hassell(L,a,b,N); ` ) :
print( `Hassell(20,3,5,N); ` ) :

```

**elif nargs = 1 and args[1] = HW then**

```

  print( `HW(u,v): The Hardy-Weinberg unerlying transformation witu (u,v,w), Eqs. (53a,53b,
  53c) in Edelestein-Keshet Ch. 3 using the fact that  $u+v+w=1$ . try: ` ) :
print( `HW(u,v); ` ) :

```

**elif nargs = 1 and args[1] = HWg then**

```

  print( `HWg(u,v,M): The Generalized Hardy-Weinberg unerlying transformation with (u,v),
  M is the survival matrix. Based on Ann Somalwar's HW3g(u,v,w) (only retain the first two
  components and replace w by 1-u-v) ` ) :
print( `Try: ` ) :

```



```
print( `HWg(u,v,[[1,2,1],[2,3,4],[1,3,2]]); ` ) :
```

```
elif nargs = 1 and args[1] = IsContStable then
```

```
print( `IsContStable(M): inputs a numeric matrix M (given as a list of lists M) and decides whether all its eigenvalues have real negative part. Try` ) :
```

```
print( `IsContStable([[1,-1],[-1,1]]); ` ) :
```

```
elif nargs = 1 and args[1] = IsDisStable then
```

```
print( `IsDisStable(M): inputs a numeric matrix M (given as a list of lists M) and decides whether all its eigenvalues have absolute value less than 1. Try` ) :
```

```
print( `IsDisStable([[1,-1],[-1,1]]); ` ) :
```

```
elif nargs = 1 and args[1] = JAC then
```

```
print( `JAC(F,x): The Jacobian Matrix (given as a list of lists) of the transformation F in the list of variables x. Try: ` ) :
```

```
print( `JAC([x+y,x^2+y^2],[x,y]); ` ) :
```

```
elif nargs = 1 and args[1] = Lotka then
```

```
print( `Lotka(r1,k1,r2,k2,b12,b21,N1,N2): The Lotka-Volterra continuous-time dynamical system, Eqs. (9a),(9b) (p. 224, section 6.3) of Edelstein-Keshet` ) :
```

```
print( `with populations N1, N2, and parameters r1,r2,k1,k2, b12, b21 (called there beta_12 and beta_21)` ) :
```

```
print( `Try: ` ) :
```

```
print( `Lotka(r1,k1,r2,k2,b12,b21,N1,N2); ` ) :
```

```
print( `Lotka(1,2,2,3,1,2,N1,N2); ` ) :
```

```
elif nargs = 1 and args[1] = May75 then
```

```
print( `May75(r,K,N): The discrete-time, single-species dynamical model of May (1975) given by Eq. (8) in Edelstein-Keshet section 3.1 (p. 75)` ) :
```

```
print( `where the variable is N (the population), and the parameters are r and K` ) :
```

```
print( `Try: ` ) :
```

```
print( `May75(r,K,N); ` ) :
```

```
print( `May75(3/2,2,N); ` ) :
```

```
elif nargs = 1 and args[1] = NicholsonBailey then
```

```
print( `NicholsonBailey(L,a,c): The discrete-time, double-species dynamical model of Nicholson and Bailey (1935), given by Eqs. (21a)(21b) in Edelstein-Keshet section 3.2 (p. 81)` ) :
```

```
print( `where the variables are N (hosts) and parasites (P) and the parameters are L (called Lambda there), a, and c` ) :
```

```
print( `Try: ` ) :
```

```
print( `NicholsonBailey(L,a,c,N,P); ` ) :
```

```
print( `NicholsonBailey(2,0.068,1,N,P); ` ) :
```

**elif nargs = 1 and args[1] = NicholsonBaileyM then**

```
print( `NicholsonBaileyM(a,r,K,N,B): The discrete-time, double-species dynamical model of
the MODIFIED Nicholson and Bailey model (1935), given by Eqs. (28a)(28b) in Edelstein-
Keshet section 3.4 (p. 84)` ) :
print( `where the variables are N (hosts) and parasites (P) and the parameters are r and K` ) :
print( `Try: ` ) :
print( `NicholsonBaileyM(r,a,K,N,P); ` ) :
print( `plot(OrbF(NicholsonBaileyM(0.5,0.11,15,N,P),[N,P],[3.,5.],1,1000),style=point); ` ) :
```

**elif nargs = 1 and args[1] = Orb then**

```
print( `Orb(F,x,x0,K1,K2): Inputs a transformation F in the list of variables x with initial
point pt, outputs the trajectory of ` ) :
print( `of the discrete dynamical system (i.e. solutions of the difference equation): x(n)=F(x
(n-1)) with x(0)=x0 from n=K1 to n=K2. ` ) :
print( `For the full trajectory (from n=0 to n=K2), use K1=0. Try: ` ) :
print( `Orb([5/2*x*(1-x)],[x], [0.5], 1000,1010); ` ) :
print( `Orb([(1+x+y)/(2+x+y),(6+x+y)/(2+4*x+5*y)],[x,y], [2.,3.], 1000,1010); ` ) :
```

**elif nargs = 1 and args[1] = OrbF then**

```
print( `OrbF(F,x,x0,K1,K2): Same as Orb(F,x,x0,K1,K2) but in floating-point ` ) :
print( `Inputs a transformation F in the list of variables x with initial point pt, outputs the
trajectory ` ) :
print( `of the discrete dynamical system (i.e. solutions of the difference equation): x(n)=F(x
(n-1)) with x(0)=x0 from n=K1 to n=K2. ` ) :
print( `For the full trajectory (from n=0 to n=K2), use K1=0. Try: ` ) :
print( `OrbF([5/2*x*(1-x)],[x], [0.5], 1000,1010); ` ) :
print( `OrbF(AllenSIR(1,1/3,1/3,x,y),[x,y],[2.,3.],1000,1010); ` ) :
```

**elif nargs = 1 and args[1] = OrbK then**

```
print( `Orbk(k,z,f,INI,K1,K2): Given a positive integer k, a letter (symbol), z, an expression f
of z[1], ..., z[k] (representing a multi-variable function of the variables z[1],...,z[k]` ) :
print( `a vector INI representing the initial values [x[1],..., x[k]], and (in applications)
positive integres K1 and K2, outputs the ` ) :
print( `values of the sequence starting at n=K1 and ending at n=K2. of the sequence
satisfying the difference equation` ) :
print( `x[n]=f(x[n-1],x[n-2],..., x[n-k+1]): ` ) :
print( `This is a generalization to higher-order difference equation of procedure Orb(f,x,x0,
K1,K2). For example, try: ` ) :
print( `Orbk(1,z,5/2*z[1]*(1-z[1]),[0.5],1000,1010); ` ) :
print( `To get the Fibonacci sequence, type: ` ) :
print( `Orbk(2,z,z[1]+z[2],[1,1],1000,1010); ` ) :
print( `` ) :
print( `To get the part of the orbit between n=1000 and n=1010, of the 3rd order recurrence
given in Eq. (4) of the Ladas-Amleh paper ` ) :
print( `https://sites.math.rutgers.edu/~zeilberg/Bio21/AmlehLadas.pdf` ) :
print( `with initial conditions x(1)=1, x(2)=3, x(3)=5, Type: ` ) :
```

```
print( `Orbk(3,z,z[2]/(z[2]+z[3]),[1.,3.,5.],1000,1010);` ) :
```

```
print( `` ) :
```

```
print( `To get the part of the orbit between n=1000 and n=1010, of the 3rd order recurrence  
given in Eq. (5) of the Ladas-Amleh paper` ) :
```

```
print( `with initial conditions x(0)=1, x(1)=3, x(2)=5, Type: ` ) :
```

```
print( `Orbk(3,z,(z[1]+z[3])/z[2],[1.,3.,5.],1000,1010);` ) :
```

```
print( `` ) :
```

```
print( `To get the part of the orbit between n=1000 and n=1010, of the 3rd order recurrence  
given in Eq. (6) of the Ladas-Amleh paper` ) :
```

```
print( `with initial conditions x(0)=1, x(1)=3, x(2)=5, Type: ` ) :
```

```
print( `Orbk(3,z,(1+z[3])/z[1],[1.,3.,5.],1000,1010);` ) :
```

```
print( `` ) :
```

```
print( `To get the part of the orbit between n=1000 and n=1010, of the 3rd order recurrence  
given in Eq. (7) of the Ladas-Amleh paper` ) :
```

```
print( `with initial conditions x(0)=1, x(1)=3, x(2)=5, Type: ` ) :
```

```
print( `Orbk(3,z,(1+z[1])/(z[2]+z[3]),[1.,3.,5.],1000,1010);` ) :
```

```
elif nargs = 1 and args[1] = OrbkF then
```

```
print( `OrbkF(k,z,f,INI,K1,K2): Same as Orbk(k,z,f,INI,K1,K2), but in floating-point (to get  
around Maple's annoying habit not to automatically convert to floating point exp  
(floatingpoint)` ) :
```

```
print( `To investigate the long-term behavior Linda Allen's Conjecture 2 of` ) :
```

```
print( `https://sites.math.rutgers.edu/~zeilberg/Bio21/AllenSIR.pdf` ) :
```

```
print( `with initial conditions x(0)=0.3, x(1)=0.4, a=3, b=2 Type: ` ) :
```

```
print( `a:=0.3; b:=0.2; OrbkF(2,z,z[1]*(1-b) + (1-z[1])*(1-exp(-a*z[2])),[0.3,0.4],1000,  
1010);` ) :
```

```
print( `then type ` ) :
```

```
print( `solve(b*y-(1-y)*(1-exp(-a*y)),y);` ) :
```

```
elif nargs = 1 and args[1] = PhaseDiag then
```

```
print( `PhaseDiag(F,x,pt,h,A): Inputs a transformation F in the list of variables x (of length  
2), i.e. a mapping from R^2 to R^2 gives the` ) :
```

```
print( `The phase diagram of the solution with initial condition x(0)=pt` ) :
```

```
print( `dx/dt=F[1](x(t)) by a discrete time dynamical system with step-size h from t=0 to t=A` ) :
```

```
print( `Try: ` ) :
```

```
print( `PhaseDiag([x*(1-y),y*(1-x)],[x,y],[0.5,0.5], 0.01, 10);` ) :
```

```
elif nargs = 1 and args[1] = PhaseDiagE then
```

```
print( `PhaseDiagE(F,x,pt,h,A): Inputs a transformation F in the list of variables x (of length  
2), i.e. a mapping from R^2 to R^2 gives the` ) :
```

```

print( `The phase diagram of the solution with initial condition  $x(0)=pt$ ` ) :
print( `dx/dt=F[1](x(t)) using dsolve. It should only be used for linear system` ) :
print( `Try: ` ) :
print( `PhaseDiagE([y,-x],[x,y],[0,1],10); ` ) :

```

**elif nargs = 1 and args[1] = RandNice then**

```

print( `RandNice(x,K): A random transformation in the set of variables x where each
component is a product of two affine-linear expressions.` ) :
print( `To generate examples of continuous time dynamical systems` ) :
print( `Try: RandNice([x,y],100); ` ) :

```

**elif nargs = 1 and args[1] = RT then**

```

print( `RT(var,K): A random rational transformation of numerator and denominator degrees
1 from  $R^k$  to  $R^k$  (where  $k=nops(var)$ , with positive integer coefficients from 1 to K The
inputs are a list of variables x and a positive integer K` ) :
print( `is for generating examples. Try: ` ) :
print( `RT([x,y],10); ` ) :

```

**elif nargs = 1 and args[1] = SEquP then**

```

print( `SEquP(F,x): Given a transformation F in the list of variables finds all the Stable
Equilibrium points of the continuous-time dynamical system  $x'(t)=F(x(t))$ ` ) :
print( `SEquP([5/2*x*(1-x)],[x]); ` ) :
print( `SEquP([y*(1-x-y),x*(3-2*x-y)],[x,y]); ` ) :

```

**elif nargs = 1 and args[1] = SFP then**

```

print( `SFP(F,x): Given a transformation F in the list of variables finds all the STABLE fixed
point of the transformation  $x \rightarrow F(x)$ , i.e. the set of solutions of` ) :
print( `the system  $\{x[1]=F[1], \dots, x[k]=F[k]\}$  that are stable. Try: ` ) :
print( `SFP([5/2*x*(1-x)],[x]); ` ) :
print( `SFP([(1+x+y)/(2+3*x+y), (3+x+2*y)/(5+x+3*y)],[x,y]); ` ) :

```

**elif nargs = 1 and args[1] = SIRS then**

```

print( `SIRS(s,i,beta,gamma,nu,N): The SIRS dynamical model with parameters beta,gamma,
nu,N (see section 6.6 of Edelstein-Keshet), s is the number of` ) :
print( `Susceptibles, i is the number of infected, (the number of removed is given by N-s-i). N
is the total population. Try: ` ) :
print( `SIRS(s,i,beta,gamma,nu,N); ` ) :

```

**elif nargs = 1 and args[1] = SIRSdemo then**

```

print( `SIRSdemo(N,IN,gamma,nu,h,A): A demonstration of the SIRS model with NUMBERS
N: The total population, IN: The number of infected individuals at the start` ) :
print( `parameters gamma, and nu and various beta changing from  $0.1*(nu/N)$  to  $4*(nu/N)$ .
Using a discretization with mesh size h and going until  $t=A$ .` ) :
print( `Try: ` ) :
print( `SIRSdemo(1000,200,1,1,0.01,10); ` ) :

```

```

elif nargs = 1 and args[1] = TimeSeries then
print( `TimeSeries(F,x,pt,h,A,i): Inputs a transformation F in the list of variables x` ) :
    print( `The time-series of x[i] vs. time of the Dynamical system approximating the the
    autonomous continuous dynamical process` ) :
print( `dx/dt=F(x(t)) by a discrete time dynamical system with step-size h from t=0 to t=A` ) :
print( `Try: ` ) :
print( `TimeSeries([x*(1-y),y*(1-x)],[x,y],[0.5,0.5], 0.01, 10,1);` ) :

```

```

elif nargs = 1 and args[1] = TimeSeriesE then
print( `TimeSeriesE(F,x,pt,A,i): Inputs a transformation F in the list of variables x, outputs` ) :
    print( `The time-series of x[i] vs. time of the Dynamical system using the EXACT solutions via
    dsolve (note that it is usuall not possible)` ) :
    print( `It works for linear transformations, and is a good check with the approximate
    TimeSeries(F,x,pt,h,A,i) that uses discretization with` ) :
print( `dx/dt=F(x(t)) by a discrete time dynamical system with step-size h from t=0 to t=A` ) :
print( `Try: ` ) :
print( `TimeSeriesE([y,-x],[x,y],[1,0], 10,1);` ) :

```

```

elif nargs = 1 and args[1] = ToSys then
    print( `ToSys(k,z,f): converts the kth order difference equation  $x(n)=f(x[n-1],x[n-2],\dots,x[n-k])$ 
    to a first-order system` ) :
    print( ` $x1(n)=F(x1(n-1),x2(n-1), \dots,xk(n-1))$ , it gives the unerlying transormation, followed by
    the set of variables` ) :
print( `Try: ` ) :
print( `ToSys(2,z,z[1] + z[2]);` ) :

```

```

elif nargs = 1 and args[1] = Valery then
    print( `Valery(L,a,b,N): The discrete-time, single-species dynamical model of Valery,
    Gradwell, and Hassel (1973) given by Eq. (2) in Edelstein-Keshet section 3.1 (p. 74)` ) :
    print( `where the variable is N (the population), and the parameters are L (called Lambda
    there), is the reproduction rate, and a (called alpha there) and b` ) :
    print( `are the other two parameters such that  $1/(a*N^{(-b)})$  is the faction of the population
    that survives. L,a,b, can be symbolic or numeric` ) :
print( `Try: ` ) :
print( `Valery(L,a,b,N);` ) :
print( `Valery(3,2,1,N);` ) :

```

```

elif nargs = 1 and args[1] = Volterra then
    print( `Volterra(a,b,c,d,x,y): The (simple, original) Volterra predator-prey continuous-time
    dynamical system with parameters a,b,c,d` ) :
print( ` Given by Eqs. (7a) (7b) in Edelstein-Keshet p. 219 (section 6.2).` ) :
print( `a,b,c,d may be symbolic or numeric` ) :
print( `Try: ` ) :
print( `Volterra(a,b,c,d,x,y);` ) :
print( `Volterra(1,2,3,4,x,y);` ) :

```

```

elif nargs = 1 and args[1] = VolterraM then

```

```

    print( `VolterraM(a,b,c,d,x,K,y): The MODIFIED Volterra predator-prey continuous-time
    dynamical system with parameters a,b,c,d,K` ) :
print( ` Given by Eqs. (8a) (8b) in Edelstein-Keshet p. 220 (section 6.2). ` ) :
print( `a,b,c,d ,K may be symbolic or numeric` ) :
print( `Try: ` ) :
print( `VolterraM(a,b,c,d,K,x,y);` ) :
print( `VolterraM(1,2,3,4,3,x,y);` ) :

```

**else**

```
print( `There is no such thing as`, args ) :
```

**fi:**

**end:**

*#Orb(F,x,x0,K1,K2): Inputs a transformation F in the list of variables x with initial point pt, outputs the trajectory*

*#of the discrete dynamical system (i.e. solutions of the difference equation):  $x(n)=F(x(n-1))$  with  $x(0)=x0$  from  $n=K1$  to  $n=K2$ .*

*#For the full trajectory (from  $n=0$  to  $n=K2$ ), use  $K1=0$ . Try:*

*#Orb(5/2\*x\*(1-x),[x], [0.5], 1000,1010);*

*#Orb((1+x+y)/(2+x+y),[x,y], [2.,3.], 1000,1010);*

*Orb :=proc(F, x, x0, K1, K2) local x1, i, L, i1, i2 :*

```

if not (type(F, list) and type(x, list) and type(x0, list) and nops(F) = nops(x) and nops(x)
    = nops(x0) and type(K1, integer) and type(K2, integer) and K1 ≥ 0 and K1 ≤ K2) then
    print( `bad input` ) :
    RETURN(FAIL) :

```

**fi:**

```
x1 := x0 :
```

**for i from 0 to K1 - 1 do**

```
x1 := [seq(subs( {seq(x[i2] = x1[i2], i2 = 1 ..nops(x))}, F[i1]), i1 = 1 ..nops(F)) ] :
```

**od:**

```
L := [x1] :
```

**for i from K1 to K2 - 1 do**

```
x1 := [seq(subs( {seq(x[i2] = x1[i2], i2 = 1 ..nops(x))}, F[i1]), i1 = 1 ..nops(F)) ] :
```

```
L := [op(L), expand(x1)] : #we append it to the list
```

**od:**

*L : #that's the output*

**end:**

*#OrbF(F,x,x0,K1,K2): Inputs a transformation F in the list of variables x with initial point pt, outputs the trajectory in FLOATING-POINT*

*#of the discrete dynamical system (i.e. solutions of the difference equation):  $x(n)=F(x(n-1))$  with  $x(0)=x0$  from  $n=K1$  to  $n=K2$ .*

*#For the full trajectory (from  $n=0$  to  $n=K2$ ), use  $K1=0$ . Try:*

*#OrbF(5/2\*x\*(1-x),[x], [0.5], 1000,1010);*

*#OrbF((1+x+y)/(2+x+y),[x,y], [2.,3.], 1000,1010);*

*OrbF :=proc(F, x, x0, K1, K2) local x1, i, L, i1, i2 :*

**if not** (type(F, list) **and** type(x, list) **and** type(x0, list) **and** nops(F) = nops(x) **and** nops(x) = nops(x0) **and** type(K1, integer) **and** type(K2, integer) **and**  $K1 \geq 0$  **and**  $K1 \leq K2$ ) **then**

*print( `bad input` ) :*

*RETURN(FAIL) :*

**fi:**

*x1 := evalf(x0) :*

**for** i **from** 0 **to** K1-1 **do**

*x1 := [seq(evalf(subs( {seq(x[i2]=x1[i2], i2=1..nops(x))}, F[i1])), i1=1..nops(F))]:*

**od:**

*L := [x1] :*

**for** i **from** K1 **to** K2-1 **do**

*x1 := [seq(evalf(subs( {seq(x[i2]=x1[i2], i2=1..nops(x))}, F[i1])), i1=1..nops(F))]:*

*L := [op(L), expand(x1)] : #we append it to the list*

**od:**

*L : #that's the output*

**end:**

*#FP(F,x): Given a transformation F in the list of variables finds all the fixed point of the transformation  $x \rightarrow F(x)$ , i.e. the set of solutions of*

*#the system  $\{x[1]=F[1], \dots, x[k]=F[k]\}$ . Try:*

*#FP([5/2\*x\*(1-x),[x]]);*

*#FP([(1+x+y)/(2+3\*x+y), (3+x+2\*y)/(5+x+3\*y)],[x,y]);*

*FP :=proc(F, x) local i, sol :*

**if not** (type(F, list) **and** type(x, list) **and** nops(F) = nops(x)) **then**

*print( `bad input` ) :*

*RETURN(FAIL) :*

**fi:**

$sol := \{solve(\{seq(F[i]=x[i], i=1 ..nops(F))\}, \{op(x)\}, allsolutions = true)\} :$

$\{seq(subs(sol[i], x), i=1 ..nops(sol))\} :$

**end:**

*#RT(var,K): A random rational transformation of numerator and denominator degrees 1 from  $R^k$  to  $R^k$  (where  $k=nops(var)$ ), with positive integer coefficients from 1 to K The inputs are a list of variables x and a positive integer K*

*#is for generating examples*

*#Try:*

*#RT([x,y],10);*

*RT := proc(x, K) local ra, i, i1 :*

*if not (type(x, list) and type(K, integer) and K > 0) then*

*print( `bad input` ) :*

*RETURN(FAIL) :*

**fi:**

*ra := rand(1 ..K) : #random integer from -K to K*

*[seq((ra() + add(ra() \* x[i1], i1 = 1 ..nops(x))) / (ra() + add(ra() \* x[i1], i1 = 1 ..nops(x))), i = 1 ..nops(x)) ] :*

**end:**

*#IsContStable(M): inputs a numeric matrix M (given as a list of lists M) and decides whether all its eigenvalues have real negative part. Try*

*#IsContStable(Matrix([[1,-1],[-1,1]]));*

*IsContStable := proc(M) local Ei1, i :*

*#k:=nops(M):*

*Ei1 := Eigenvalues(evalf(Matrix(M))) :*

*evalb(max(seq(coeff(Ei1[i], 1, 0), i = 1 ..nops(M))) < 0):*

**end:**

*#IsDisStable(M): inputs a numeric matrix M (given as a list of lists M) and decides whether all its eigenvalues have absolute value less than 1*

*#IsDisStable(Matrix([[1,-1],[-1,1]]));*

*IsDisStable := proc(M) local Ei1, i :*

*Ei1 := Eigenvalues(evalf(Matrix(M))) :*

*evalb(max(seq(abs(Ei1[i]), i = 1 ..nops(M))) < 1):*

**end:**



*#JAC(F,x): The Jacobian Matrix (given as a list of lists) of the transformation F in the list of variables x. Try:*  
*#JAC([x+y,x^2+y^2],[x,y]):*

```
JAC :=proc(F, x) local i, j :
if not (type(F, list) and type(x, list) and nops(F) = nops(x)) then
  print( `Bad input` ) :
  RETURN(FAIL) :
fi:

normal([seq([seq(diff(F[i], x[j]), j = 1 ..nops(x))], i = 1 ..nops(F))]) :

end:
```

*#SFP(F,x): Given a transformation F in the list of variables finds all the STABLE fixed point of the transformation  $x \rightarrow F(x)$ , i.e. the set of solutions of the system  $\{x[1]=F[1], \dots, x[k]=F[k]\}$  that are stable. Try:*

```
#SFP([5/2*x*(1-x)],[x]);
#SFP([(1+x+y)/(2+3*x+y), (3+x+2*y)/(5+x+3*y)],[x,y]);
SFP :=proc(F, x) local i, Fi, St, J, J0, pt :
if not (type(F, list) and type(x, list) and nops(F) = nops(x)) then
  print( `bad input` ) :
  RETURN(FAIL) :
fi:
Fi := evalf(FP(F, x)) : #Fi is the set of fixed points in floating-point
```

*St := { } : #St is the set of stable fixed points, that starts out empty*

*J := JAC(F, x) : #The general Jacobian in terms of the list of variables x*

```
for pt in Fi do #we examine each fixed point, one at a time
  J0 := subs({seq(x[i]=pt[i], i = 1 ..nops(x))}, J) :
  #J0 is the NUMETRICAL Jacobian matrix evaluated at the examined fixed point
```

```
if IsDisStable(J0) then
  St := St union{pt} : #if it is stable we include it
fi:
```

**od**:

*St : #The output is the set of all the successful fixed points that happened to be stable*  
**end**:

*#Orbk(k,z,f,INI,K1,K2): Given a positive integer k, a letter (symbol), z, an expression f of z [1], ..., z[k] (representing a multi-variable function of the variables z[1],...,z[k]*

```

# a vector INI representing the initial values [x[1],..., x[k]], and (in applications) positive
integers K1 and K2, outputs the

# values of the sequence starting at n=K1 and ending at n=K2. of the sequence satisfying the
difference equation
##x[n]=f(x[n-1],x[n-2],..., x[n-k+1]):

# This is a generalization to higher-order difference equation of procedure Orb(f,x,x0,K1,K2)
. For example
#Orbk(1,z,5/2*z[1]*(1-z[1]),[0.5],1000,1010); should be the same as
#Orb(5/2*z[1]*(1-z[1]),z[1],[0.5],1000,1010);
#Try:
#Orbk(2,z,(5/4)*z[1]-(3/8)*z[2],[1,2],1000,1010);

Orbk := proc(k, z, f, INI, K1, K2) local L, i, newguy :
L := INI: #We start out with the list of initial values

if not (type(k, integer) and type(z, symbol) and type(INI, list) and nops(INI) = k and type(K1,
integer) and type(K2, integer) and K1 ≥ 1 and K2 ≥ K1) then
#checking that the input is OK
print( `bad input` ) :
RETURN(FAIL) :
fi:

while nops(L) < K2 do
newguy := subs( {seq(z[i]=L[-i], i=1..k)}, f) :
#Using what we know about the value yesterday, the day before yesterday, ... up to k days
before yesterday we find the value of the sequence today
L := [op(L), newguy] : #we append the new value to the running list of values of our sequence
od:

[op(K1..K2, L)]:

end:

#OrbkF(k,z,f,INI,K1,K2): Like Orbk(k,z,f,INI,K1,K2) but in floating-point
#OrbkF(1,z,5/2*z[1]*(1-z[1]),[0.5],1000,1010); should be the same as
#OrbkF(5/2*z[1]*(1-z[1]),z[1],[0.5],1000,1010);
#Try:
#OrbkF(2,z,(5/4)*z[1]-(3/8)*z[2],[1,2],1000,1010);

OrbkF := proc(k, z, f, INI, K1, K2) local L, i, newguy :
L := INI: #We start out with the list of initial values

if not (type(k, integer) and type(z, symbol) and type(INI, list) and nops(INI) = k and type(K1,
integer) and type(K2, integer) and K1 > 0 and K2 > K1) then
#checking that the input is OK

```

```

print( `bad input` ) :
RETURN(FAIL) :
fi:

while nops(L) < K2 do
newguy := evalf( subs( { seq(z[i]=L[-i], i=1..k) }, f ) ) :
    #Using what we know about the value yesterday, the day before yesterday, ... up to k days
    before yesterday we find the value of the sequence today
L := [op(L), newguy] : #we append the new value to the running list of values of our sequence
od:

```

```
[op(K1..K2, L)]:
```

```
end:
```

*#ToSys(k,z,f): converts the kth order difference equation  $x(n)=f(x[n-1],x[n-2],\dots,x[n-k])$  to a first-order system*

*#x1(n)=F(x1(n-1),x2(n-1), ...,xk(n-1)), it gives the unerlying transformation, followed by the set of variables*

```
#x2(n)=x1(n-1)
```

```
#Try:
```

```
#ToSys(2,z,z[1]+z[2]);
```

```
ToSys := proc(k, z, f) local i :
```

```
[f, seq(z[i-1], i=2..k) ], [seq(z[i], i=1..k) ]:
```

```
end:
```

*#HW3(u,v,w): The Hardy-Weinberg unerlying transformation witu (u,v,w), Eqs. (53a,53b, 53c) in Edelestein-Keshet Ch. 3*

```
HW3 := proc(u, v, w) : [u^2 + u * v + (1/4) * v^2, u * v + 2 * u * w + 1/2 * v^2 + v * w, 1/4 * v^2 + v * w + w^2] : end:
```

*#HW(u,v): The Hardy-Weinberg unerlying transformation witu (u,v,w), Eqs. (53a,53b,53c) in Edelestein-Keshet Ch. 3 using the fact that  $u+v+w=1$*

```
HW := proc(u, v) : expand([u^2 + u * v + (1/4) * v^2, u * v + 2 * u * (1-u-v) + 1/2 * v^2 + v * (1-u-v) ]), [u, v] : end:
```

*#HW3g(u,v,w,M): The Hardy-Weinberg unerlying transformation with (u,v,w),*

GENERALIZED Eqs. with the 3 by 3 matrix  $M$  (53a,53b,53c) in Edelestein-Keshet Ch. 3  
 #Based on Anne Somalwar's solution of the bonus problem from hw15, see the end of  
 #from <https://sites.math.rutgers.edu/~zeilberg/Bio21/HW15posted/hw15AnneSomalwar.pdf>  
 HW3g :=**proc**( $u, v, w, M$ ) **local** tot, LI :  
 LI := [

$M[1][1] * u^2 + (M[1][2] + M[2][1]) / 2 * u * v + M[2][2] * (1/4) * v^2,$

$(M[1][2] + M[2][1]) / 2 * u * v + (M[1][3] + M[3][1]) * u * w + M[2][2] / 2 * v^2$   
 $+ (M[2][3] + M[3][2]) / 2 * v * w,$

$M[2][2] * 1/4 * v^2 + (M[2][3] + M[3][2]) / 2 * v * w + M[3][3] * w^2$  :

tot := LI[1] + LI[2] + LI[3] :

[LI[1]/tot, LI[2]/tot, LI[3]/tot] :

**end**:

#HWg( $u,v,M$ ): The Generalized Hardy-Weinberg underlying transformation with ( $u,v$ ),  $M$  is the survival matrix. Based on Ann Somalwar's HW3g( $u,v,w$ ) (only retain the first two components and replace  $w$  by  $1-u-v$ )

HWg :=**proc**( $u, v, M$ ) **local** LI, w :

LI := HW3g( $u, v, w, M$ ) :

normal(subs( $w = 1 - u - v$ , [LI[1], LI[2]])) :

**end**:

#RandNice( $x,K$ ): A random transformation in the set of variables  $x$  where each component is a product of two affine-linear expressions.

#To generate examples

#Try: RandNice( $[x,y],100$ );

RandNice :=**proc**( $x, K$ ) **local** ra, i :

ra := rand(1..K) :

[seq((ra() - add(ra() \*  $x[i]$ ,  $i = 1 .. nops(x)$ )) \* (ra() - add(ra() \*  $x[i]$ ,  $i = 1 .. nops(x)$ )),  $i = 1 .. nops(x)$ )] :

**end**:

#EquP( $F,x$ ): Given a transformation  $F$  in the list of variables finds all the Equilibrium points of the continuous-time dynamical system  $x'(t)=F(x(t))$

#EquP( $[5/2*x*(1-x),[x]]$ );

#EquP( $[y*(1-x-y),x*(3-2*x-y)],[x,y]$ );

EquP :=**proc**( $F, x$ ) **local** i, sol :

**if not** (type( $F, list$ ) **and** type( $x, list$ ) **and** nops( $F$ ) = nops( $x$ )) **then**

```

print( `bad input` ) :
RETURN(FAIL) :
fi:

sol := {solve( {op(F)}, {op(x)}, allsolutions = true) } :

{seq(subs(sol[i], x), i = 1 ..nops(sol)) } :

```

**end:**

```

#SEquP(F,x): Given a transformation F in the list of variables x describing the
CONTINUOUS-time dynamical system  $x'(t)=F(x(t))$ 
#Finds the set of Stable Equilibria. Try:
#SEquP([y*(1-x-y),x*(3-2*x-y)], [x,y]);
SEquP := proc(F, x) local i, Fi, St, J, J0, pt :
if not (type(F, list) and type(x, list) and nops(F) = nops(x)) then
print( `bad input` ) :
RETURN(FAIL) :
fi:
Fi := evalf(EquP(F, x)) : #Fi is the set of equilibrium points in floating-point

St := { } : #St is the set of stable fixed points, that starts out empty

J := JAC(F, x) : #The general Jacobian in terms of the list of variables x

for pt in Fi do #we examine each fixed point, one at a time
J0 := subs( {seq(x[i] = pt[i], i = 1 ..nops(x)) }, J) :
#J0 is the NUMETRICAL Jacobian matrix evaluated at the examined fixed point

if IsContStable(J0) then
St := St union {pt} : #if it is stable we include it
fi:

od:

St : #The output is the set of all the successful fixed points that happened to be stable
end:

```

*#Dis(F,x,pt,h,A): Inputs a transformation F in the list of variables x*

```

#The approximate orbit of the Dynamical system approximating the the autonomous
continuous dynamical process
#dx/dt=F[1](x(t)) by a discrete time dynamical system with step-size h from t=0 to t=A
#Try:
#Dis([x*(1-y),y*(1-x)], [x,y], [0.5,0.5], 0.01, 10);
Dis := proc(F, x, pt, h, A) local L, i :

```

```

if not (type(F, list) and type(x, list) and type(pt, list) and nops(F) = nops(x) and nops(F)
    = nops(pt) and type(h, numeric) and h ≤ 0.1 and type(A, numeric) and A > 0) then
    print( `bad input` ) :
    RETURN(FAIL) :
fi:

```

```

L := Orb( [seq(x[i] + h * F[i], i = 1 ..nops(F)) ], x, pt, 0, trunc(A/h) ) :

```

```

L := [seq([i * h, L[i]], i = 1 ..nops(L)) ] :

```

```

end:

```

*#SIRS(s,i,beta,gamma,nu,N): The SIRS dynamical model with parameters beta,gamma, nu,N (see section 6.6 of Edelstein-Keshet), s is the number of*

*#Susceptibles, i is the number of infected, (the number of removed is given by N-s-i). N is the total population*

```

SIRS :=proc(s, i, beta, gamma, nu, N) : [-beta * s * i + gamma * (N-s-i), beta * s * i - nu * i] :
end:

```

*#SIRSDemo(N,IN,gamma,nu,h,A): A demonstration of the SIRS model with NUMBERS N: The total population, IN: The number of infected individuals at the start*

*#parameters gamma, and nu and various beta changing from 0.1\*(nu/N) to 4\*(nu/N) . Using a discretization with mesh size h and going until t=A.*

```

#Try:

```

```

#SIRSDemo(1000,200,1,1,0.01,10);

```

```

SIRSDemo :=proc(N, IN, gamma, nu, h, A) local s, i, L, beta, i1 :

```

```

    print( `This is a numerical demonstration of the R0 phenomenon in the SIRS model using
    discretization with mesh size=`, h, `and letting it run until time t=`, A ) :

```

```

    print( `with population size`, N, `and fixed parameters nu=`, nu, `and gamma=`, gamma ) :

```

```

    print( `where we change beta from 0.2*nu/N to 4*nu/N` ) :

```

```

    print( `Recall that the epidemic will persist if beta exceeds nu/N, that in this case is`, nu/N ) :

```

```

    print( `We start with`, IN, `infected individuals, 0 removed and hence`, N-IN, `susceptible` ) :

```

```

    print( `We will show what happens once time is close to`, A ) :

```

```

for i1 from 1 by 2 to 40 do

```

```

    beta := i1/10 * (nu/N) :

```

```

    print( `beta is`, i1/10, `times the threshold value` ) :

```

```

    L := Dis(SIRS(s, i, beta, gamma, nu, N), [s, i], [N-IN, IN], h, A) :

```

```

    print( `the long-term behavior is` ) :

```

```

    print( [op(nops(L)-3 ..nops(L), L) ] ) :

```

```

od:

```

**end:**

*#TimeSeries(F,x,pt,h,A,i): Inputs a transformation F in the list of variables x*

*#The time-series of  $x[i]$  vs. time of the Dynamical system approximating the the autonomous continuous dynamical process*

*#dx/dt=F[1](x(t)) by a discrete time dynamical system with step-size h from t=0 to t=A*

*#Try:*

*#TimeSeries([x\*(1-y),y\*(1-x)],[x,y],[0.5,0.5], 0.01, 10,1);*

*TimeSeries :=proc(F, x, pt, h, A, i) local L, i1 :*

**if not** (type(F, list) **and** type(x, list) **and** type(pt, list) **and** nops(F) = nops(x) **and** nops(F) = nops(pt) **and** type(h, numeric) **and** h ≤ 0.1 **and** type(A, numeric) **and** A > 0 **and** 1 ≤ i **and** i ≤ nops(x) ) **then**

*print('bad input') :*

*RETURN(FAIL) :*

**fi:**

*L := Dis(F, x, pt, h, A) :*

*plot([seq([L[i1][1], L[i1][2]][i], i1 = 1 ..nops(L) ])) :*

**end:**

*#PhaseDiag(F,x,pt,h,A): Inputs a transformation F in the list of variables x (of length 2), i.e. a mapping from  $R^2$  to  $R^2$  gives the*

*#The phase diagram of the solution with initial condition  $x(0)=pt$*

*#dx/dt=F[1](x(t)) by a discrete time dynamical system with step-size h from t=0 to t=A*

*#Try:*

*#PhaseDiag([x\*(1-y),y\*(1-x)],[x,y],[0.5,0.5], 0.01, 10);*

*PhaseDiag :=proc(F, x, pt, h, A) local L, i1 :*

**if not** (type(F, list) **and** type(x, list) **and** type(pt, list) **and** nops(F) = nops(x) **and** nops(F) = nops(pt) **and** nops(x) = 2 **and** type(h, numeric) **and** h ≤ 0.1 **and** type(A, numeric) **and** A > 0) **then**

*print('bad input') :*

*RETURN(FAIL) :*

**fi:**

*L := Dis(F, x, pt, h, A) :*

*plot([seq(L[i1][2], i1 = 1 ..nops(L) ]), style = point) :*

**end:**

*#ComK(F,x,K): inputs a transformation F in the list of variables x, outputs the composition of*

*F with itself K times. Try:*  
 #ComK([k\*x\*(1-x)],[x],2);  
 #ComK([x\*(1-y),y\*(1-x)],[x,y],4);

```
ComK :=proc(F, x, K) local F1, i :
option remember :
if K = 0 then
RETURN(x) :
elif K = 1 then
RETURN(F) :
else
F1 := ComK(F, x, K-1) :
RETURN(normal(subs({seq(x[i] = F[i], i = 1 ..nops(x))}, F1))) :
fi:

end:
```

*#AllenSIR(a,b,c,x,y): The Linda Allen discrete SIR model given in <https://sites.math.rutgers.edu/~zeilberg/Bio21/AllenSIR.pdf>*  
*#with parameters a,b,c. try:*  
 #AllenSIR(1,1/3,1/3,x,y);  
 AllenSIR :=proc(a, b, c, x, y)  
 [x\*(1-b-c) + y\*(1-exp(-a\*x)), (1-y)\*b + y\*exp(-a\*x)]:  
 end:

*#AllenSIRg(a,b,c,alpha,beta,x,y): The GENERALIZED Linda Allen discrete SIR model given in <https://sites.math.rutgers.edu/~zeilberg/Bio21/AllenSIR.pdf>*  
*#with parameters a,b,c. Try:*  
*#where the exponents of x\_n and y\_n are alpha and beta. Note that*  
*#AllenSIRg(a,b,c,1,1,x,y) is the same as AllenSIR(a,b,c,x,y): Try:*  
 #AllenSIRg(1,1/3,1/3,1.2,1.2,x,y);  
 AllenSIRg :=proc(a, b, c, alpha, beta, x, y)  
 [x^alpha\*(1-b-c) + y^beta\*(1-exp(-a\*x)), (1-y^beta)\*b + y^beta\*exp(-a\*x)]:  
 end:

*#TimeSeriesE(F,x,x0,A,i): Inputs a transformation F in the list of variables x, outputs*

*#The time-series of x[i] vs. time of the Dynamical system using the exact solutions via dsolve (note that it is usually not possible)*

*#It works for linear transformations, and is a good check with the approximate TimeSeries(F, x,x0,h,A,i) that uses discretization with*  
*#dx/dt=F[1](x(t)) by a discrete time dynamical system with step-size h from t=0 to t=A*  
 #Try:  
 #TimeSeriesE([y,-x],[x,y],[0,1], 10,1);



```

TimeSeriesE := proc(F, x, x0, A, i) local sol, t, i1, F1 :
if not (type(F, list) and type(x, list) and type(x0, list) and nops(F) = nops(x) and nops(F)
    = nops(x0) and type(A, numeric) and A > 0 and 1 ≤ i and i ≤ nops(x) ) then
    print( `bad input` ) :
    RETURN(FAIL) :
fi:

F1 := subs( {seq(x[i1] = X[i1](t), i1 = 1 ..nops(x))}, F) :
sol := dsolve( {seq(diff(X[i1](t), t) = F1[i1], i1 = 1 ..nops(x)), seq(X[i1](0) = x0[i1], i1 = 1
    ..nops(x0))} ) :

plot(subs(sol, X[i](t)), t = 0 ..A) :

end:

```

*#PhaseDiagE(F,x,x0,A): Inputs a transformation F in the PAIR of variables x, outputs*

*#The Phase diagram [x[1],x[2]] (forgetting about time, that becomes a parameter) of the Dynamical system using the exact solutions via dsolve (note that it is usuall not possible)*

*#It works for linear transformations, and is a good check with the approximate TimeSeries(F, x,x0,h,A,i)*

*#Try:*

*#TimeSeriesE([y,-x],[x,y],[0,1], 10);*

*PhaseDiagE :=proc(F, x, x0, A) local sol, t, i1, X, F1 :*

```

if not (type(F, list) and type(x, list) and nops(x) = 2 and type(x0, list) and nops(F) = nops(x)
    and nops(F) = nops(x0) and type(A, numeric) and A > 0 ) then
    print( `bad input` ) :
    RETURN(FAIL) :
fi:

```

```

F1 := subs( {seq(x[i1] = X[i1](t), i1 = 1 ..nops(x))}, F) :
sol := dsolve( {seq(diff(X[i1](t), t) = F1[i1], i1 = 1 ..nops(x)), seq(X[i1](0) = x0[i1], i1 = 1
    ..nops(x0))} ) :

plot([subs(sol, X[1](t)), subs(sol, X[2](t)), t = 0 ..A]) :

```

**end**:

*#ChemoStat(N,C,a1,a2): The Chemostat continuous-time dynamical system with N=Bacterial population density, and C=nutrient Concentration in growth chamber (see Table 4.1 of Edelstein-Keshet, p. 122)*

*#with paramerts a1, a2, Equations (19a\_, (19b) in Edelestein-Keshet p. 127 (section 4.5, where they are called alpa1, alpha2). a1 and a2 can be symbolic or numeric. Try:*

```
#ChemoStat(N,C,a1,a2);
```

```
#ChemoStat(N,C,2,3);
```

```
ChemoStat := proc(N, C, a1, a2) :  
[a1 * C / (1 + C) * N - N, -C / (1 + C) * N - C + a2] :  
end;
```

```
#Volterra(a,b,c,d,x,y): The (simple, original) Volterra predator-prey continuous-time  
dynamical system with parameters a,b,c,d
```

```
#Eqs. (7a) (7b) in Edelstein-Keshet p. 219 (section 6.2)
```

```
#a,b,c,d may be symbolic or numeric
```

```
#Try:
```

```
#Volterra(a,b,c,d,x,y);
```

```
#Volterra(1,2,3,4,x,y);
```

```
Volterra := proc(a, b, c, d, x, y)
```

```
[a * x - b * x * y, -c * y + d * x * y] :
```

```
end;
```

```
#VolterraM(a,b,c,d,K,x,y): The modified Volterra predator-prey continuous-time dynamical  
system with parameters a,b,c,d,K
```

```
#Eqs. (8a) (8b) in Edelstein-Keshet p. 220 (section 6.2)
```

```
#a,b,c,d,K may be symbolic or numeric
```

```
#Try:
```

```
#VolterraM(a,b,c,d,K,x,y);
```

```
#VolterraM(1,2,3,4,2,x,y);
```

```
VolterraM := proc(a, b, c, K, d, x, y)
```

```
[a * x * (1 - x / K) - b * x * y, -c * y + d * x * y] :
```

```
end;
```

```
#Lotka(r1,k1,r2,k2,b12,b21,N1,N2): The Lotka-Volterra continuous-time dynamical system,  
Eqs. (9a),(9b) (p. 224, section 6.3) of Edelstein-Keshet
```

```
#with populations N1, N2, and parameters r1,r2,k1,k2, b12, b21 (called there beta_12 and  
beta_21)
```

```
#Try:
```

```
#Lotka(r1,k1,r2,k2,b12,b21,N1,N2);
```

```
#Lotka(1,2,2,3,1,2,N1,N2);
```

```
Lotka := proc(r1, k1, r2, k2, b12, b21, N1, N2) :
```

```
[r1 * N1 * (k1 - N1 - b12 * N2) / k1, r2 * N2 * (k2 - N2 - b21 * N1) / k2] :
```

```
end;
```

```

#GeneNet(a0,a,b,n,m1,m2,m3,p1,p2,p3): The continuous-time dynamical system, with
quantities m1,m2,m3,p1,p2,p3, due to M. Elowitz and S. Leibler
#described in the Ellner-Guckenheimer book, Eq. (4.1) (chapter 4, p. 112)
#and parameters a0 (called alpha_0 there), a (called alpha there), b (called beta there) and n. Try:
#GeneNet(0,0.5,0.2,2,m1,m2,m3,p1,p2,p3);
GeneNet := proc(a0, a, b, n, m1, m2, m3, p1, p2, p3) :
[ -m1 + a / (1 + p3^n) + a0, -m2 + a / (1 + p1^n) + a0, -m3 + a / (1 + p2^n) + a0, -b
* (p1 - m1), -b * (p2 - m2), -b * (p3 - m3) ] :
end:

```

```

#Valery(L,a,b,N): The discrete-time, single-species dynamical model of Valery, Gradwell,
and Hassel (1973) given by Eq. (2) in Edelstein-Keshet section 3.1 (p. 74)

```

```

#where the variable is N (the population), and the parameters are L (called Lambda there), is
the reproduction rate, and a (called alpha there) and b

```

```

#are the other two parameters such that  $1/(a*N^{(-b)})$  is the fraction of the population that
survives. L,a,b, can be symbolic or numeric

```

```

#Try:
#Valery(L,a,b,N);
#Valery(3,2,1,N);
Valery := proc(L, a, b, N) :
[ (L/a) * N^(1-b) ] :
end:

```

```

#May75(r,K,N): The discrete-time, single-species dynamical model of May (1975) given by
Eq. (8) in Edelstein-Keshet section 3.1 (p. 75)

```

```

#where the variable is N (the population), and the parameters are r and K

```

```

#Try:
#May75(r,K,N);
#May75(3/2,2,N);
May75 := proc(r, K, N) :
[ N * exp(r * (1 - N/K)) ] :
end:

```

```

#Hassell(L,a,b,N): The discrete-time, single-species dynamical model of Hassell (1975) given
by Eq. (13) in Edelstein-Keshet section 3.1 (p. 75)

```

```

#where the variable is N (the population), and the parameters are L (called Lambda there), a,
and b

```

```

#Try:

```

```

#Hassell(L,a,b,N);
#Hassell(20,3,5,N);
Hassell := proc(L, a, b, N) :
[L * N * (1 + a * N)^(-b)]:
end:

```

*#NicholsonBailey(L,a,c): The discrete-time, double-species dynamical model of Nicholson and Bailey (1935), given by Eqs. (21a)(21b) in Edelstein-Keshet section 3.2 (p. 81)*

*#where the variables are N (hosts) and parasites (P) and the parameters are L (called Lambda there), a, and c*

```

#Try:
#NicholsonBailey(L,a,c,N,P);
#NicholsonBailey(2,0.068,1,N,P);
NicholsonBailey := proc(L, a, c, N, P)
[L * N * exp(-a * P), c * N * (1 - exp(-a * P))]:
end:

```

*#NicholsonBaileyM(a,r,K,N,B): The discrete-time, double-species dynamical model of the MODIFIED Nicholson and Bailey model (1935), given by Eqs. (28a)(28b) in Edelstein-Keshet section 3.4 (p. 84)*

*#where the variables are N (hosts) and parasites (P) and the parameters are r and K*

```

#Try:
#NicholsonBaileyM(r,a,K,N,P);
#NicholsonBaileyM(0.5,0.1,14,N,P);
NicholsonBaileyM := proc(r, a, K, N, P)
[N * exp(r * (1 - N/K) - a * P), N * (1 - exp(-a * P))]:
end:

```

*First Written: Nov. 2021*

*This is DMB.txt, A Maple package to explore Dynamical models in Biology (both discrete and continuous)*

*accompanying the class Dynamical Models in Biology, Rutgers University. Taught by Dr. Z. (Doron Zeilberger)*

*The most current version is available on WWW at:  
<http://sites.math.rutgers.edu/~zeilberg/tokhniot/DMB.txt> .*

Please report all bugs to: DoronZeil at gmail dot com .

For general help, and a list of the MAIN functions,  
type "Help():". For specific help type "Help(procedure\_name);"

-----  
For a list of the supporting functions type: Help1();  
For help with any of them type: Help(ProcedureName);

-----  
For a list of the functions that give examples of Discrete-time dynamical systems (some famous),  
type: HelpDDM());

For help with any of them type: Help(ProcedureName);

-----  
For a list of the functions continuous-time dynamical systems (some famous) type: HelpCDM());

For help with any of them type: Help(ProcedureName);

(1)

```
> #Nicholas DiMarzio final exam
> #1
#Set up the recurrence  $R(n)=2 \cdot R(n-1)-R(n-3)$ 
> #Using a modified version of my proc from hw1 I find the sol
> R := proc(n, c0, c1, c2) option remember;
  if n = 0 then
    c0;
  elif n = 1 then
    c1 :
  elif n = 2 then
    c2 :
  else
    expand( $2 \cdot R(n - 1, c0, c1, c2) - R(n - 3, c0, c1, c2)$ ) : #Recurrence
  fi:
end:
```

```
#Procedure should be able to calculate the given recurrence scenario
```

```
> R(3, 1, 1, 2)
```

3

(2)

```
> R(4, 1, 1, 2)
```

5

(3)

```
> R(5, 1, 1, 2)
```

(4)

```
> f1 := R(1000, 1, 1, 2)
```

```
f1 :=
```

```
70330367711422815821835254877183549770181269836358732742604905087154537118\
19693357974224949456261173348775044924176599108818636326545022364710601205\
3374121273867339111198139373125598767690091902245245323403501
```

```
> f2 := R(999, 1, 1, 2)
```

```
f2 :=
```

```
43466557686937456435688527675040625802564660517371780402481729089536555417\
94905189040387984007925516929592259308032263477520968962323987332247116164\
2996440906533187938298969649928516003704476137795166849228875
```

```
> evalf( $\left(\frac{f1}{f2}\right)$ )
```

1.618033989

```
> #Thus we find our solution
```

```
> #2
```

```
> #This is a continuous-time first order dynamical system
```

```
F :=  $\frac{5}{2}x \cdot (1 - x) \cdot \left(1 - \frac{1}{2}x\right)$ ; #Underlying Function
```

$$F := \frac{5x(1-x)\left(1 - \frac{x}{2}\right)}{2}$$

```
> EquP([F], [x]); #2a all equilibrium solutions
```

{[0], [1], [2]}

```
> SEquP([F], [x]) #2b all stable equilibrium solutions
```

{[1.]}

```
> #2c
```

```
> G := dsolve( $\left\{\left\{D(x)(t) = \frac{5}{2} \cdot x(t) \cdot (1 - x(t)) \cdot \left(1 - \frac{1}{2} \cdot x(t)\right), x(0) = 0.1\right\}, x(t)\right\}$ )
```

$$G := x(t) = \frac{19 e^{\frac{5t}{2}}}{81 \left(-\frac{19 e^{\frac{5t}{2}}}{81} - 1\right) \left(-\frac{1}{\sqrt{1 + \frac{19 e^{\frac{5t}{2}}}{81}}} - 1\right)}$$

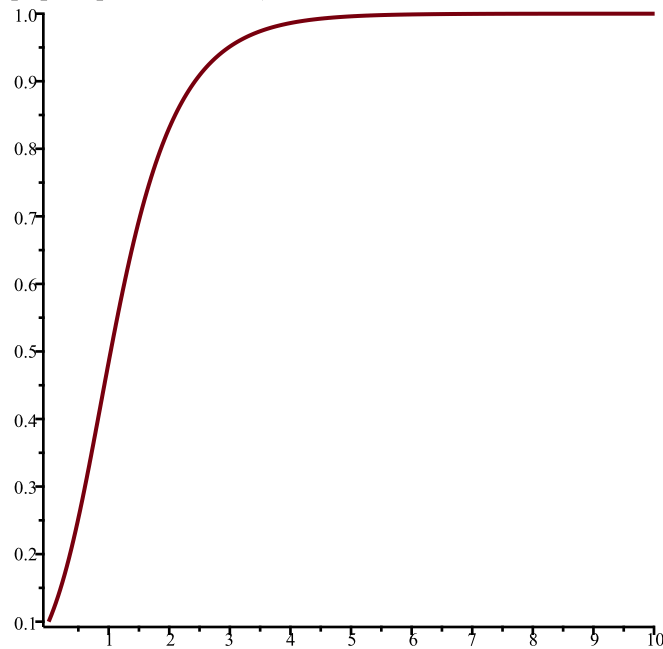
```
> evalf(eval(G, t = 100))
```

x(100) = 0.9999999999

```
> #We get very close to our stable equilibrium solution
```

```
> #We can also use time series
```

> TimeSeries([F], [x], [0.1], 0.01, 10, 1)



> #We get a horizontal asymptote at our equilibrium solution

> #3

> #This is a discrete-time dynamical system

>  $x1 := \frac{5}{2} \cdot x \cdot (1 - x) \cdot \left(1 - \frac{1}{2} \cdot x\right)$  #Underlying function

$$x1 := \frac{5x(1-x)\left(1 - \frac{x}{2}\right)}{2} \quad (13)$$

> evalf(FP([x1], [x])); #3a All Equilibrium solutions (fixed points)

$$\{[0.], [0.475304923], [2.524695077]\} \quad (14)$$

> SFP([x1], [x]); #3b All Stable Equilibrium solutions

$$\{[0.475304923]\} \quad (15)$$

> #3c) We can use Orb to see the long term behavior in this case at day 1000

> Orb([x1], [x], [0.1], 1000, 1000)[1]

$$[0.4753049232] \quad (16)$$

> #Thus we get very close to our equilibrium solution

> #4a

> F := HW3(u, v, w)

$$F := \left[ u^2 + v u + \frac{1}{4} v^2, v u + 2 u w + \frac{1}{2} v^2 + v w, \frac{1}{4} v^2 + v w + w^2 \right] \quad (17)$$

>  $x1 := u^2 + v \cdot u + \frac{1}{4} \cdot v^2$

$$x1 := u^2 + v u + \frac{1}{4} v^2 \quad (18)$$

$$\begin{aligned} > x2 := v \cdot u + 2 \cdot u \cdot w + \frac{1}{2} \cdot v^2 + v \cdot w \\ & \qquad \qquad \qquad x2 := v u + 2 u w + \frac{1}{2} v^2 + v w \end{aligned} \tag{19}$$

$$\begin{aligned} > x3 := \frac{1}{4} \cdot v^2 + v \cdot w + w^2 \\ & \qquad \qquad \qquad x3 := \frac{1}{4} v^2 + v w + w^2 \end{aligned} \tag{20}$$

$$\begin{aligned} > x1 + x2 + x3 \\ & \qquad \qquad \qquad u^2 + 2 v u + 2 u w + v^2 + 2 v w + w^2 \end{aligned} \tag{21}$$

$$\begin{aligned} > \text{factor}(\%) \\ & \qquad \qquad \qquad (v + u + w)^2 \end{aligned} \tag{22}$$

> #Thus we notice the equation is normalized

$$\begin{aligned} > \text{Orb}\left(F, [u, v, w], \left[\frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right], 0, 3\right) \\ & \qquad \qquad \qquad \left[\left[\frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right], \left[\frac{1}{4}, \frac{1}{2}, \frac{1}{4}\right], \left[\frac{1}{4}, \frac{1}{2}, \frac{1}{4}\right], \left[\frac{1}{4}, \frac{1}{2}, \frac{1}{4}\right]\right] \end{aligned} \tag{23}$$

> #It stabilizes after 1 generation

> #Thus our second generation with genotype Aa is

$$> \# \frac{1}{2}$$

> #4b)

$$\begin{aligned} > \text{Orb}\left(F, [u, v, w], \left[\frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right], 1000, 1000\right)[1] \\ & \qquad \qquad \qquad \left[\frac{1}{4}, \frac{1}{2}, \frac{1}{4}\right] \end{aligned} \tag{24}$$

> #Thus the proportion after 1000 generations is still

$$> \# \frac{1}{2}$$

> #5a

$$\begin{aligned} > G := \text{normal}(\text{HW3g}(u, v, w, [[0.2, 0.4, 0.2], [0.2, 0.2, 0.2], [0.2, 0.2, 0.2]])) \\ G := & \left[ (0.2000000000 u^2 + 0.3000000000 v u + 0.05000000000 v^2) / (0.2000000000 u^2 \right. \\ & + 0.6000000000 v u + 0.2000000000 v^2 + 0.4000000000 u w + 0.4000000000 v w \\ & + 0.2000000000 w^2), (0.3000000000 v u + 0.4000000000 u w + 0.1000000000 v^2 \\ & \left. + 0.2000000000 v w) / (0.2000000000 u^2 + 0.6000000000 v u + 0.2000000000 v^2 \right. \end{aligned} \tag{25}$$



$$+ 0.4000000000 u w + 0.4000000000 v w + 0.2000000000 w^2), (0.0500000000 v^2 + 0.2000000000 v w + 0.2000000000 w^2) / (0.2000000000 u^2 + 0.6000000000 v u + 0.2000000000 v^2 + 0.4000000000 u w + 0.4000000000 v w + 0.2000000000 w^2) ]$$

$$\begin{aligned} &> \text{Orb}\left(G, [u, v, w], \left[\frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right], 0, 2\right) \\ & \left[ \left[\frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right], [0.2750000001, 0.5000000002, 0.2250000001], [0.3027472528, \right. \\ & \left. 0.4989010991, 0.1983516484] \right] \end{aligned} \tag{26}$$

> #The second generation has proportion of 0.5000000002

> #5b

$$\begin{aligned} &> \text{SFP}(G, [u, v, w]) \\ & \quad \{[0., 0., 1.], [0.5512669098, 0.3974661804, 0.05126690980]\} \end{aligned} \tag{27}$$

$$\begin{aligned} &> \text{Orb}\left(G, [u, v, w], \left[\frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right], 1000, 1000\right)[1] \\ & \quad [0.5512669092, 0.3974661807, 0.05126690995] \end{aligned} \tag{28}$$

> #After 1000 generations the proportion is 0.3974661807

> #6

$$\begin{aligned} &> f2 := \left[ \frac{(1 + x + y)}{(2 + x + 3 \cdot y)}, \frac{(1 + x + 3 \cdot y)}{(3 + x + 2 \cdot y)} \right]; \# \text{Underlying function discrete time} \\ & \quad f2 := \left[ \frac{1 + x + y}{2 + x + 3 y}, \frac{1 + x + 3 y}{3 + x + 2 y} \right] \end{aligned} \tag{29}$$

$$\begin{aligned} &> \text{evalf}(\text{FP}(f2, [x, y])) \\ & \quad \{[0.4705902280, 0.7478789082]\} \end{aligned} \tag{30}$$

$$\begin{aligned} &> \text{evalf}(\text{SFP}(f2, [x, y])) \\ & \quad \{[0.4705902280, 0.7478789082]\} \end{aligned} \tag{31}$$

$$\begin{aligned} &> \text{evalf}(\text{Orb}(f2, [x, y], [100, 1000], 1, 10)) \\ & \quad [[0.3549323017, 1.474560152], [0.4174146768, 0.9166504659], [0.4516933984, \\ & \quad 0.7936758311], [0.4646180236, 0.7606046077], [0.4688200954, 0.7514162981], \\ & \quad [0.4700834136, 0.7488594278], [0.4704477412, 0.7481500864], [0.4705505458, \\ & \quad 0.7479538049], [0.4705792301, 0.7478995784], [0.4705871874, 0.7478846105]] \end{aligned} \tag{32}$$

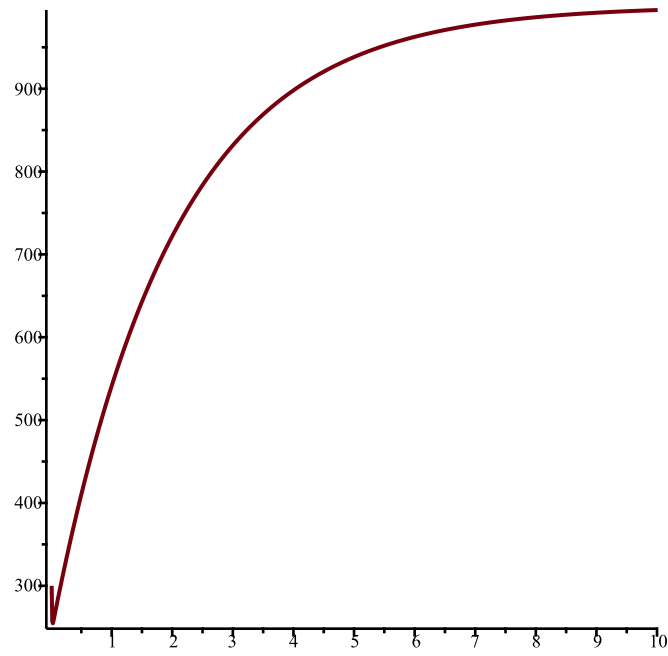
> #In the long run, y will eventually stabilize to 0.7478789082

> #7

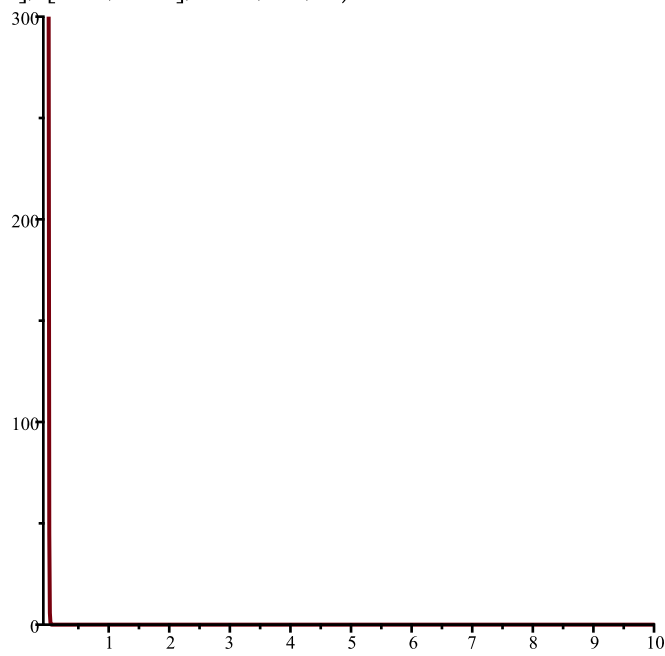
$$\begin{aligned} &> s1 := \text{SIRS}(s, i, 0.05, 0.5, 100, 1000) \\ & \quad s1 := [-0.05 s i + 500.0 - 0.5 s - 0.5 i, 0.05 s i - 100 i] \end{aligned} \tag{33}$$

$$\begin{aligned} &> \text{SEquP}(s1, [s, i]) \\ & \quad \{[1000., 0.]\} \end{aligned} \tag{34}$$

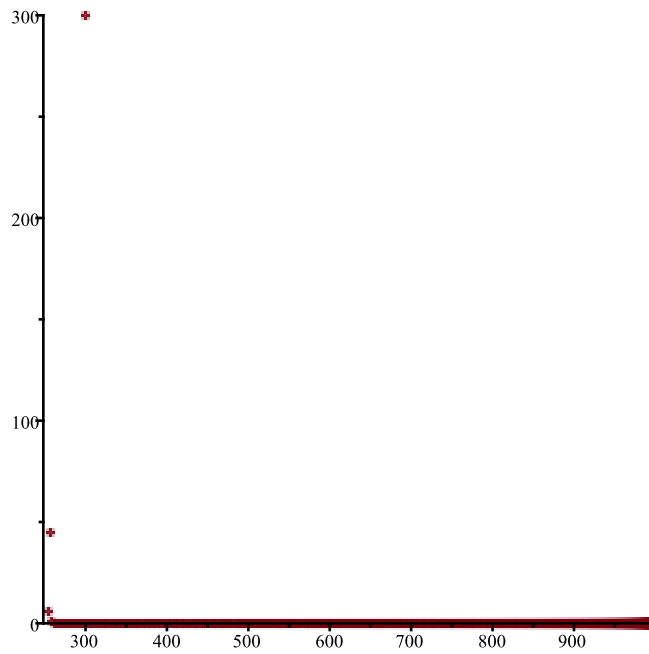
$$\begin{aligned} &> \text{TimeSeries}(s1, [s, i], [300, 300], 0.01, 10, 1) \end{aligned}$$



> *TimeSeries*(s1, [s, i], [300, 300], 0.01, 10, 2)



> *PhaseDiag*(s1, [s, i], [300, 300], 0.01, 10)



```

> #We see that the epidemic will die out
> #our equilibrium point tells us that we will have everyone survive
> #So our removed will remain at 400

```

```

> #R0:
> #  $\frac{N \cdot \beta}{\nu}$ 

```

```

> #If  $R0 < 1$  then the epidemic will die out

```

```

>  $R0 = \frac{1000 \cdot 0.05}{100}$ 

```

$$R0 = 0.5000000000$$

(35)

```

> #Thus the epidemic will die out

```

```

> #7b

```

```

> s2 := SIRS(s, i, 1.4, 0.5, 100, 1000)

```

$$s2 := [-1.4 \ s \ i + 500.0 - 0.5 \ s - 0.5 \ i, 1.4 \ s \ i - 100 \ i]$$

(36)

```

> SEquP(s2, [s, i])

```

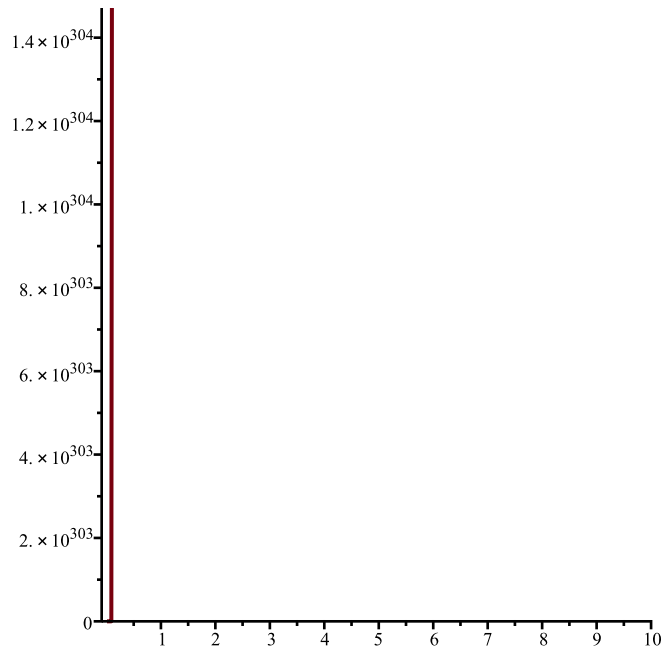
$$\{[71.42857143, 4.619758351]\}$$

(37)

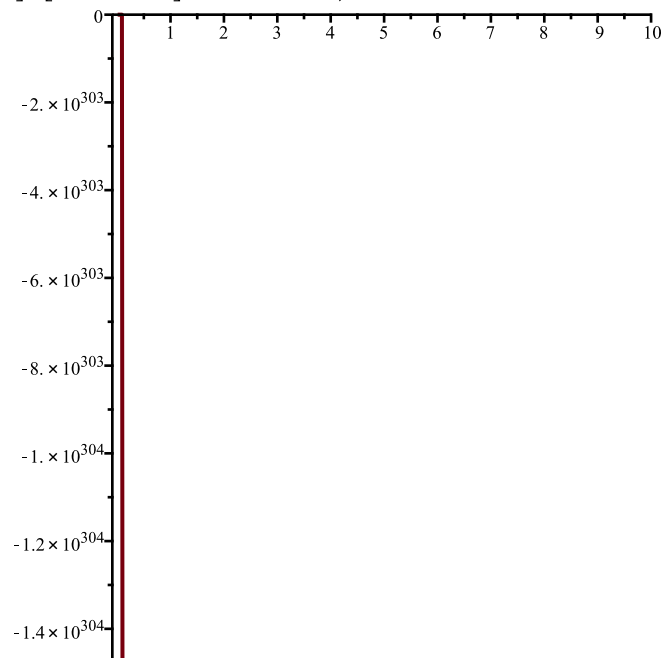
```

> TimeSeries(s2, [s, i], [300, 300], 0.01, 10, 1)

```



> TimeSeries(s2, [s, i], [300, 300], 0.01, 10, 2)



> #From our Stable Equilibrium solution,

> 71.42857143 + 2.629708600

74.05828003

(38)

> 1000 - %

925.9417200

(39)

> #Thus the removed is about 926 of the population

> #Since  $R_0 = 14 > 1$  the epidemic will persist.

> #7c

> #The cutoff would be when:

> #beta =  $\frac{nu}{N}$

> s3 := SIRS(s, i, 0.1, 0.5, 100, 1000)

$$s3 := [-0.1 s i + 500.0 - 0.5 s - 0.5 i, 0.1 s i - 100 i] \quad (40)$$

> SEquP(s3, [s, i])

$$\emptyset \quad (41)$$

> s4 := SIRS(s, i, 0.09, 0.5, 100, 1000)

$$s4 := [-0.09 s i + 500.0 - 0.5 s - 0.5 i, 0.09 s i - 100 i] \quad (42)$$

> s5 := SIRS(s, i, 0.11, 0.5, 100, 1000)

$$s5 := [-0.11 s i + 500.0 - 0.5 s - 0.5 i, 0.11 s i - 100 i] \quad (43)$$

> SEquP(s4, [s, i])

$$\{[1000., 0.]\} \quad (44)$$

> SEquP(s5, [s, i])

$$\{[909.0909091, 0.4522840344]\} \quad (45)$$

> #Thus we see the cutoff between having infected people and non infected people

>

> #8a)

> SEquP(GeneNet(0, 1, 0.2, 2, m1, m2, m3, p1, p2, p3), [m1, m2, m3, p1, p2, p3])

$$\{[0.6823278038, 0.6823278038, 0.6823278038, 0.6823278038, 0.6823278038, 0.6823278038]\} \quad (46)$$

> #The height of the horizontal asymptote is at 0.6823278038

> #8b)

> SEquP(GeneNet(0, 3, 0.2, 2, m1, m2, m3, p1, p2, p3), [m1, m2, m3, p1, p2, p3])

$$\{[1.213411663, 1.213411663, 1.213411663, 1.213411663, 1.213411663, 1.213411663]\} \quad (47)$$

> #The new height of the horizontal asymptote is at 1.213411663

> #8c)

> SEquP(GeneNet(0, 7.39, 0.2, 2, m1, m2, m3, p1, p2, p3), [m1, m2, m3, p1, p2, p3])

$$\{[1.777163792, 1.777163792, 1.777163792, 1.777163792, 1.777163792, 1.777163792]\} \quad (48)$$

> SEquP(GeneNet(0, 7.40, 0.2, 2, m1, m2, m3, p1, p2, p3), [m1, m2, m3, p1, p2, p3])

$$\emptyset \quad (49)$$

> #Thus we have found an alpha where a horizontal asymptote exists and when we add 0.01 we do not have a horizontal asymptote

> #9

> q1 := ChemoStat(N, C, 2.5, 2.7)

$$q1 := \left[ \frac{2.5 C N}{C + 1} - N, -\frac{C N}{C + 1} - C + 2.7 \right] \quad (50)$$

> SEquP(q1, [N, C])

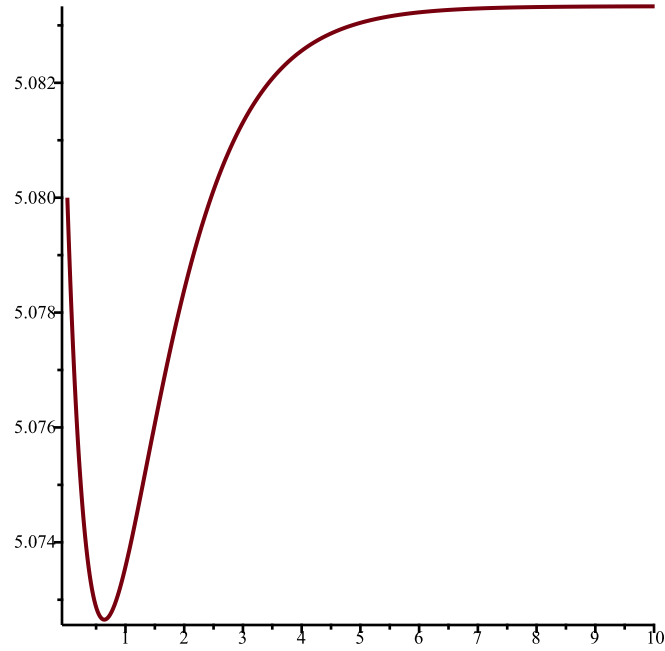
$$\{[5.083333333, 0.666666667]\} \quad (51)$$

> #N is the bacterial population and C is the Nuterient Concetration

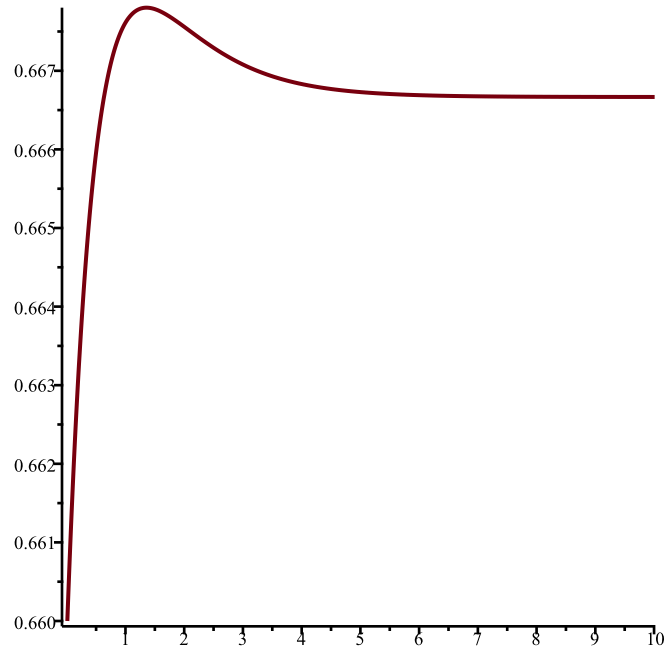
> #Thus 5.083333333 is the Bacterial population

> #0.666666667 is the nutrient concetration

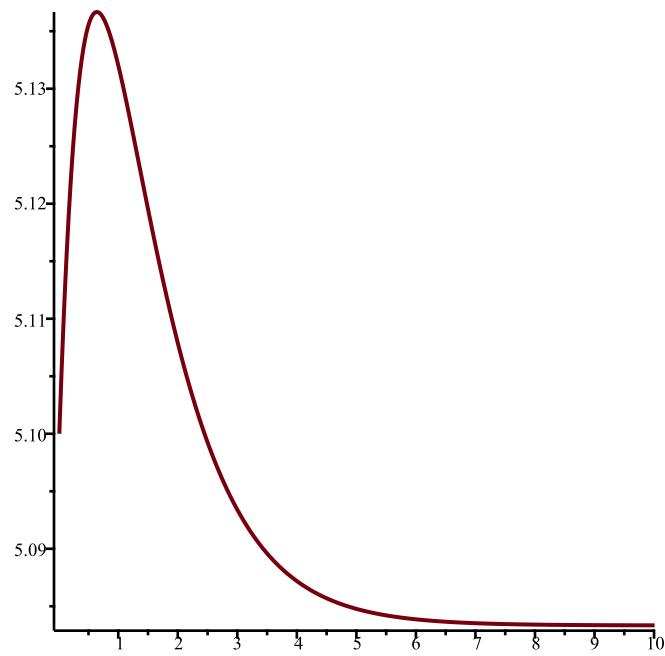
> *TimeSeries*(*q1*, [*N*, *C*], [5.08, 0.66], 0.01, 10, 1)



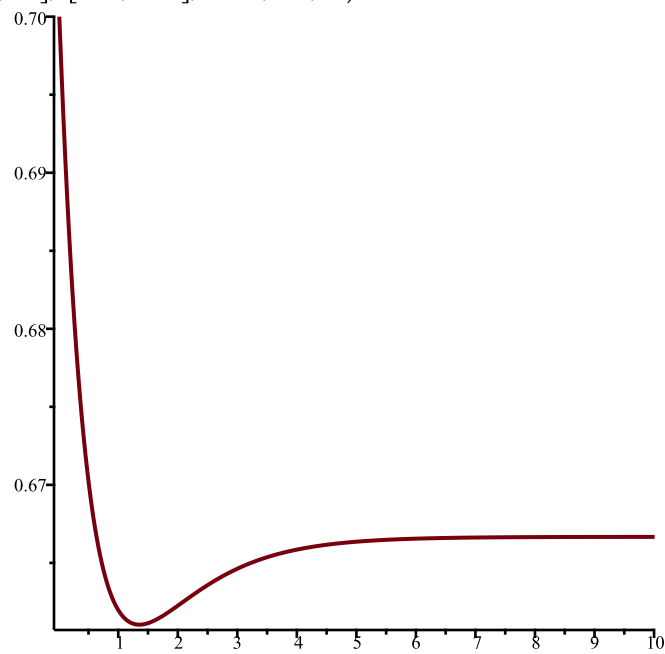
> *TimeSeries*(*q1*, [*N*, *C*], [5.08, 0.66], 0.01, 10, 2)



> *TimeSeries*(*q1*, [*N*, *C*], [5.1, 0.7], 0.01, 10, 1)

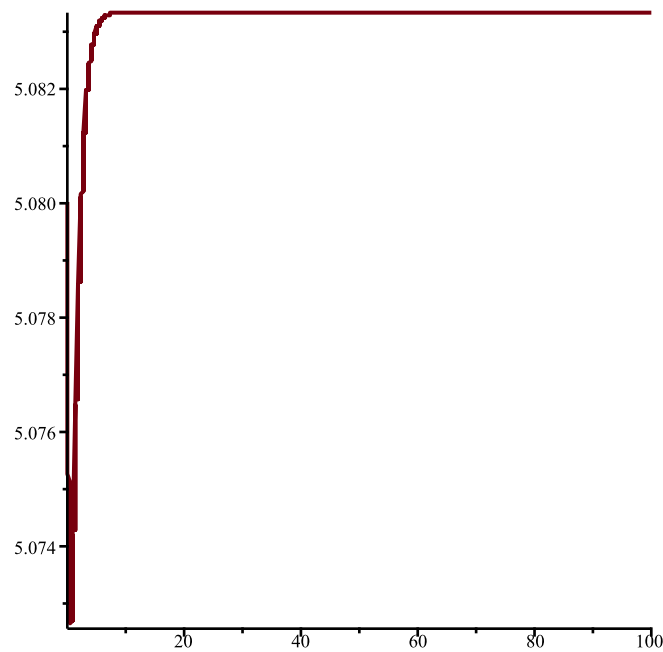


> *TimeSeries(q1, [N, C], [5.1, 0.7], 0.01, 10, 2)*

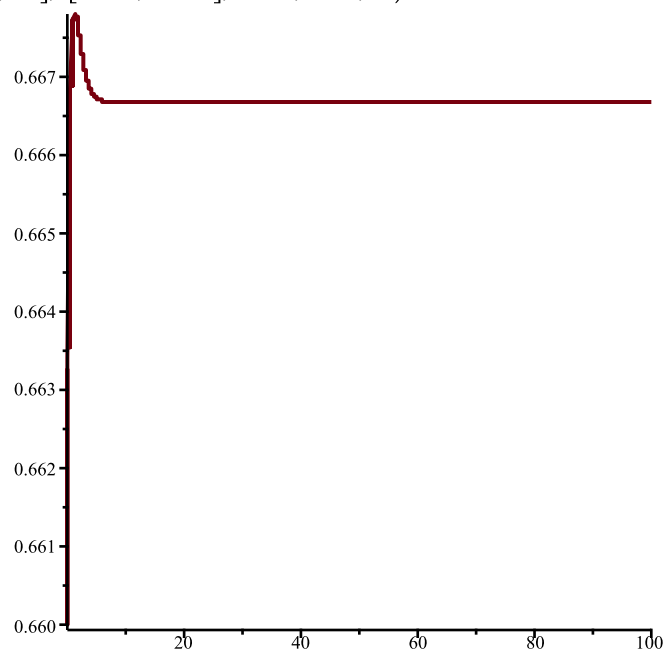


> *#Horizontal asymptotes at each stable equilibrium solution*

> *TimeSeries(q1, [N, C], [5.08, 0.66], 0.01, 100, 1)*



```
> TimeSeries(q1, [N, C], [5.08, 0.66], 0.01, 100, 2)
```



```
> #After a long time, we expect to get closer and closer to our equilibrium solutions
```

```
> #10
```

```
> #Given the probabilities, we can set up a transition matrix P
```

```
with(LinearAlgebra) :
```

```
P := evalf(Matrix([ [0.2, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1], [0.1, 0.2, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1],
[0.1, 0.1, 0.2, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1], [0.075, 0.075, 0.075, 0.4, 0.075, 0.075, 0.075, 0.075,
0.075], [0.075, 0.075, 0.075, 0.075, 0.4, 0.075, 0.075, 0.075, 0.075], [0.075, 0.075, 0.075, 0.075,
0.075, 0.4, 0.075, 0.075, 0.075], [0.05, 0.05, 0.05, 0.05, 0.05, 0.05, 0.6, 0.05, 0.05], [0.05, 0.05,
0.05, 0.05, 0.05, 0.05, 0.6, 0.05], [0.05, 0.05, 0.05, 0.05, 0.05, 0.05, 0.05, 0.6]]))
```





0.153846153846160, 0.153846153846160 ]]

> #Thus for probability that surfer will be at page 1 we get 0.0769230769230801

> #And for probability that surfer will be at page 9 we get 0.153846153846160

> #END OF DOCUMENT

>