Lotke-Volterra Equations: Population Models in Mathematica

Developed in the 1930s, the Lotke-Volterra Equations are a system of Ordinary Differential Equations (ODE) commonly used to model populations of species that are interacting with each other in a particular ecological environment. The most simple set of these are the Predator-Prey model, which is as follows:

 $\begin{aligned} x'(t) &= \alpha * x(t) - \beta * x(t) * y(t) \\ y'(t) &= -\gamma * y(t) + \delta * x(t) * y(t) \end{aligned}$

where x(t) represents the prey population as a function of time, and y(t) performs a similar role for the predator.

 α is the prey growth rate, β is the effect of predator presence on prey, γ is predator death rate and δ is the effect of prey presence on predators.

In Mathematica these equations can be implemented as such:

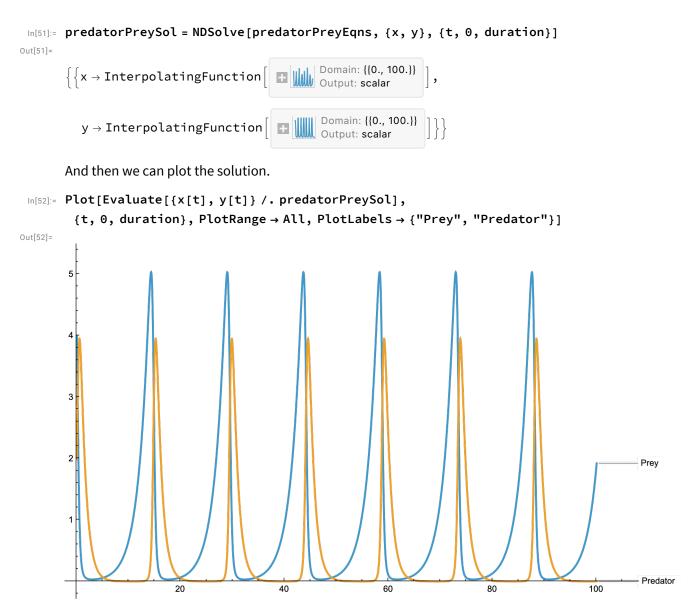
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in[42]:= (*We implement the parameters shown above first*)
a = 0.5(*prey growth rate*);
b = 1(*predator effect on prey*);
c = 1(*predator death rate*);
d = 1(*prey effect on predator*);
preyInitialPopulation = 4;
predatorInitialPopulation = 2;
duration = 100;
in[49]:= (*Then we can program the equations themselves*)
in[50]:= predatorPreyEqns = {x'[t] == a*x[t] - b*x[t]*y[t], y'[t] == -c*y[t] + d*x[t]*y[t],
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Out[50]=

 $\{x'[t] = 0.5x[t] - x[t] \times y[t], y'[t] = -y[t] + x[t] \times y[t], x[0] = 4, y[0] = 2\}$

x[0] == preyInitialPopulation, y[0] == predatorInitialPopulation}

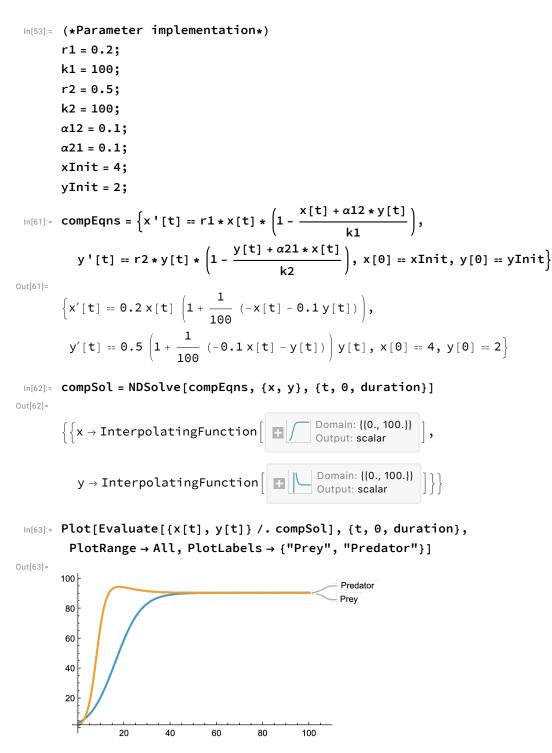
Unfortunately the Lotke-Volterra Equations don't have algebraic solutions—the best we can do is numerical analysis using NDSolve[].



The next level of complexity arises when we introduce the concept of carrying capacity—which refers to the maximum population a species can take on in the environment before overpopulation begins to decrease the number of individuals. The system (known as the Lotke-Volterra Competitive Equations) becomes this:

 $\begin{aligned} x'(t) &= r_1 \star x(t) \star \left(1 - \frac{x[t] + \alpha_{12} \, y[t]}{k_1} \right) \\ y'(t) &= r_2 \star y(t) \star \left(1 - \frac{y[t] + \alpha_{21} \, x[t]}{k_2} \right) \end{aligned}$

where r_i and k_i refer to growth and carrying capacity respectively, and $\alpha_{i,j}$ is now redefined as the effect species j has on species i (in this case, x is 1 and y is 2).



The ultimate system utilised are the Generalised Lotke-Volterra Equations, which can model n number of species. This is done using the generalised formula:

 $x_i'(t) = (r_i + A) x$

here A represents a matrix of interactions between various species. Each row of the matrix has 3 inputs,

{effect of species a on species b, effect of species b on species a, and effect of species a on itself}. An example with three species is implemented below.

```
In[64]:=
       identities = {x, y, z};
       growthRates = {3, 4, 7.2};
       initials = {0.1, 0.8, 0.3};
       interactionMatrix = {{-0.5, -1, 0}, {0, -1, -2}, {-2.6, -1.6, -3}};
       (*Creating the interaction matrix*)
       Grid[interactionMatrix] (*Visualisation*)
Out[68]=
       -0.5 - 1 0
        0 -1 -2
       -2.6 - 1.6 - 3
ln[69]:= n = 3;
       system = Table[Flatten[{identities[[i]] '[t] == (growthRates[[i]] +
                   Total[Table[interactionMatrix[i]][j] * identities[[j][t], {j, n}]) *
                identities[[i]][t], identities[[i]]'[0] == initials[[i]]}], {i, n}];
       (*Generating a system of equations from the parameters*)
       Column[system]
Out[71]=
       \{x'[t] = x[t] (3 - 0.5x[t] - y[t]), x'[0] = 0.1\}
       \{y'[t] = y[t] (4 - y[t] - 2z[t]), y'[0] = 0.8\}
       \{z'[t] = (7.2 - 2.6 x[t] - 1.6 y[t] - 3 z[t]) z[t], z'[0] = 0.3\}
In[72]:= genEq = Flatten[system]
Out[72]=
       \{x'[t] = x[t] (3 - 0.5 x[t] - y[t]), x'[0] = 0.1, y'[t] = y[t] (4 - y[t] - 2 z[t]),
        y'[0] = 0.8, z'[t] = (7.2 - 2.6 x[t] - 1.6 y[t] - 3 z[t]) z[t], z'[0] = 0.3
In[73]:= genSol = NDSolve[genEq, identities, {t, 0, duration}]
Out[73]=
                                                Domain: {{0., 100.}}
       \left\{ \left\{ x \rightarrow InterpolatingFunction \right\} \right\}
                                                                  ,
                                                Output: scalar
                                                Domain: {{0., 100.}}
         y \rightarrow InterpolatingFunction
                                        + /
                                                Output: scalar
```

Domain: {{0., 100.}}

Output: scalar

 $z \rightarrow \texttt{InterpolatingFunction}$

