# Multi-Dimensional Dynamic Systems

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## Exercise 2, Dynamic systems

Multidimensional ablalalala

library(tidyverse)
library(deSolve)
library(ggpubr)

## 1. Lotka-Volterra predator-prey equations

Lotka-Volterra (LV) predator-prey equations are given by the following system of differential equations:

$$\frac{dn}{dt} = rn - anp$$

$$\frac{dp}{dt} = anp - \mu p$$

Where n(t) is the prey population, p(t) is the predator population, r is the intrisic prey growth rate, a is the predation rate ( $attack\ rate$ ), and  $\mu$  is the predator death rate (mortality).

#### a) Write the *isoclines* of the system.

"What will the **predator** population be when the **prey** population growth equals zero?" Meaning, there is a sweet spot in predator population where the prey population is stable. This is called the *prey isocline* and *vice versa*. This value is called the *isocline* because it is the line where the population is stable.

Thus, solve each differential equation at a time (equal to zero) and we will get the corresponding population value to keep the other value at 0 growth.

$$\frac{dn}{dt} = 0 \leftrightarrow rn - anp = 0 \leftrightarrow n(r - ap) = 0$$

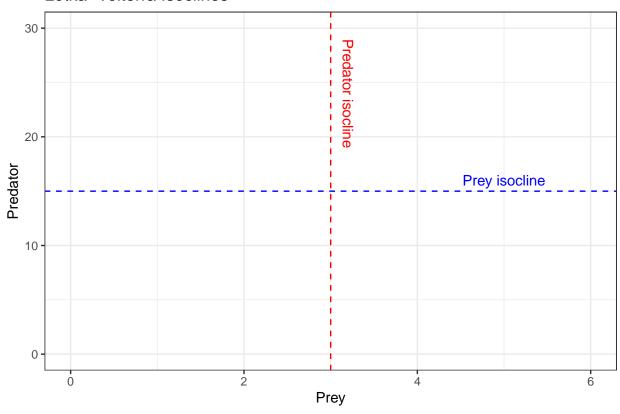
$$\frac{dp}{dt} = 0 \leftrightarrow anp - \mu p = 0 \leftrightarrow p(an - \mu) = 0$$

We get two trivial equilibrium points, n = 0, p = 0 and two equilibrium points of interest:

Prey: 
$$p = \frac{r}{a}$$
, Predator:  $n = \frac{\mu}{a}$ 

```
# Define the function for LV isoclines
LV_iso <- function(r, a, mu){
 ne <- r/a
 pe <- mu/a
 return(list(prey_iso = ne, pred_iso = pe))
# Define the parameters
P \leftarrow list(r = 3,
          a = 0.2,
          mu = 0.6
isoc <- LV_iso(P$r, P$a, P$mu)</pre>
# Plot the isoclines
# Create a blank ggplot object with the axes limits where i add the isoclines
# using horizontal and vertical lines.
ggplot() +
  geom_hline(yintercept = isoc$prey_iso, color = "blue", linetype = "dashed") +
  geom_vline(xintercept = isoc$pred_iso, color = "red", linetype = "dashed") +
 labs(x = "Prey", y = "Predator") +
  # Multiply by 2 as the isoclines are the equilibrium points.
  xlim(0, isoc$pred_iso*2) +
  ylim(0, isoc$prey_iso*2) +
  theme bw() +
  annotate("text", x = 5, y = 16, label = "Prey isocline", color = "blue") +
  annotate("text", x = 3.2, y = 24, label = "Predator isocline", color = "red",
           angle = -90) +
  ggtitle("Lotka-Volterra Isoclines")
```

## Lotka-Volterra Isoclines



#### b) Simulate using ode() and plot a time-plot and phase-plane plot.

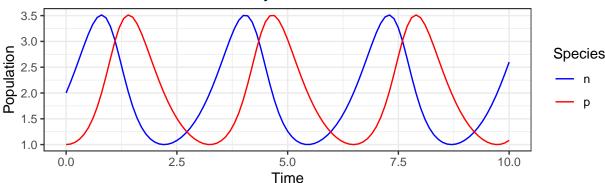
Run simulations with ode() from deSolve in R. Use two state variables: prey and predator densities. Return their time derivatives as a vector in a list. Simulate the system, outputting a matrix with time, prey density, and predator density. Plot both populations over time and create a phase-plane plot (predator vs. prey). Incorporate isoclines using LV\_iso().

- 1. Simulation Setup: Utilize the ode() function from the deSolve package.
- 2. State Variables: Focus on two variables prey and predator densities.
- 3. Output Format: System function should return a list with a vector of these time derivatives.
- 4. Run Simulation: Generate a matrix output with columns for time, prey density, and predator density.
- 5. **Plotting**: Create a time plot showing both populations. Construct a phase-plane plot with predator vs. prey densities.
- 6. **Isoclines**: Use LV\_iso() function to add isoclines to the phase-plane plot.

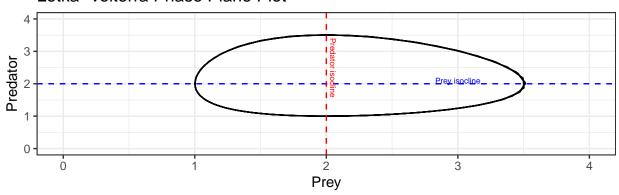
```
# Define the function for LV isoclines
LV_iso <- function(r, a, mu){
  ne <- r/a
  pe <- mu/a
  return(list(prey_iso = ne, pred_iso = pe))
}</pre>
```

```
# LV system function
LV_sys <- function(t, np, P){
  # State variables
 n \leftarrow np[1]
  p <- np[2]
  # Differential equations
  dndt <- P$r*n - P$a*n*p</pre>
 dpdt <- P$a*n*p - P$mu*p
 return(list(c(dndt, dpdt)))
# Define the parameters
P2 <- list(
 r = 2
 a = 1,
 mu = 2
  )
# Define the initial population size
np0 \leftarrow c(n = 2, p = 1)
# Time range
time_vec \leftarrow seq(0, 10, 0.1)
# Run the simulation
res_sim <- ode(y = np0, times = time_vec, func = LV_sys, parms = P2)</pre>
# Pivot the data to a long formated dataframe
df_sim <- res_sim %>%
  as.data.frame() %>%
  pivot_longer(cols = c("n", "p"), names_to = "Species",
               values_to = "Population")
# Calculate the isoclines
isoc <- LV_iso(P2\$r, P2\$a, P2\$mu)
# Plotting the results
timeplot <- df_sim %>%
  ggplot(aes(x = time, y = Population, color = Species)) +
  geom_line() +
  scale_color_manual(values = c("blue", "red")) +
  theme_bw() +
  labs(x = "Time", y = "Population") +
  ggtitle("Lotka-Volterra Predator-Prey Simulation")
# Phase plane plot
phase_plane <- df_sim %>%
  pivot_wider(names_from = Species, values_from = Population) %>%
  ggplot(aes(x = n, y = p)) +
  geom_path() +
```

## Lotka-Volterra Predator-Prey Simulation



## Lotka-Volterra Phase Plane Plot



## c) Calculate the Jacobian matrix

To determine the stability of an equilibrium, one should calculate the *Jacobian* matrix of the system at the equilibrium point. If all *eigenvalues* of the Jacobian matrix have a **negative real part**, the equilibrium is stable.

Calculate the Jacobian matrix for the system at the equilibrium point. Write a function that takes three parameters  $(r, a, \mu)$  as input and returns the Jacobian matrix as output. We have extended the LV equation with c as the conversion efficiency of prey to predator biomass:

$$\frac{dn}{dt} = f_n(n, p) = rn - anp$$

$$\frac{dp}{dt} = f_p(n, p) = canp - \mu p$$

i) Calculating the Jacobian matrix: The partial derivative of  $f_n$  with respect to n is written  $\frac{\partial f_n}{\partial n}$ . It can be interpreted as the slope of the surface parallel to the n-axis. It is calculated by 'pretending' that all other variables (p) are constant, and vice versa for  $\frac{\partial f_n}{\partial p}$ .

Thus the partial derivatives of  $f_n$  and  $f_p$  are:

$$\begin{split} \frac{\partial f_n}{\partial n} &= r - ap, \quad \frac{\partial f_n}{\partial p} = -an \\ \frac{\partial f_p}{\partial n} &= cap, \quad \frac{\partial f_p}{\partial p} = can - \mu \end{split}$$

Giving us the Jacobian matrix:

$$J = \begin{bmatrix} r - ap & -an \\ cap & can - \mu \end{bmatrix}$$

Now, with the added conversion efficiency c, the equilibrium points are:

Prey: 
$$p = \frac{r}{a}$$
, Predator:  $n = \frac{\mu}{ca}$ 

Lets test the function with these as well.

```
# LV Jacobian matrix
LV_jac <- function(np, P){
  r <- P$r
  a <- P$a
  mu <- P$mu
  c <- P$c
  n \leftarrow np[1]
  p \leftarrow np[2]
  J \leftarrow matrix(c(r-a*p, -a*n, c*a*p, c*a*n-mu), nrow = 2, ncol = 2)
  return(J)
# Define the parameters
P3 <- list(
  r = 2,
  a = 1,
  mu = 2,
  c = 1
# Define initial population
np \leftarrow c(n = 2, p = 1)
```

```
# The non-trivial equilibrium points
np_eq \leftarrow c(n = P3\$mu/(P3\$c*P3\$a), p = P3\$r/P3\$a)
# The Jacobian matrix
LV_jac(np, P3)
        [,1] [,2]
## [1,]
           1
## [2,]
          -2
\# The Jacobian matrix at the equilibrium point
LV_jac(np_eq, P3)
        [,1] [,2]
## [1,]
## [2,]
          -2
                 0
```

## d) Calculate the eigenvalues of the Jacobian matrix

Use the LV\_jac() function to calculate the Jacobian for a set of parameter values. Also calculate the cycle period associated with the complex eigenvalues. Is it a good approximation of the length of the actual cycles?

```
# The Jacobian matrix
J <- LV_jac(np_eq, P3)

eigenval <- eigen(J)

# Compare the eigenvalues to the cycle period
2*pi/Im(eigenval$values[1])</pre>
```

## [1] 3.141593