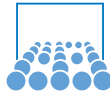


Scientific Computing I

Module 3: Population Modelling – Continuous Models (Parts III and IV)

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Winter 2015/2016



A Linear Model

First Example: Arms Race

Second Example: Competition

A Non-Linear Model

The Non-Linear Competition Model

Predator-Prey

Part IV: Analysis of ODE Models – Two Species

Direction Fields and Critical Points in 2D

Critical Points in 2D

2D Direction Fields

Summary

Analysis of Systems of ODE

Homogeneous Systems

Eigenvalues and Critical Points

Stability of Linear Systems

Stability of Non-Linear Systems

Part III

More Than One Species – Systems of ODE

A Linear Model

- similar to Verhulst's saturation model
- additional growth term proportional to other species
- leads to system of differential equations:

$$\begin{aligned}\dot{p}(t) &= b_1 + a_{11}p(t) + a_{12}q(t) \\ \dot{q}(t) &= b_2 + a_{21}p(t) + a_{22}q(t)\end{aligned}$$

- typically:
 - $b_1 > 0, b_2 > 0$ (growth term)
 - $a_{11} < 0, a_{22} < 0$ (saturation)
 - $a_{12}, a_{21}?$

First Example: Arms Race

- armament of two (hostile) countries
- our suspicion: $a_{12} > 0, a_{21} > 0$

Observation:

- long-time behaviour depends on size of parameters
- steady-state solutions exist
- solutions exist that show unlimited growth

Second Example: Competition

- two species sharing a common natural habitat
- competition: $a_{12} < 0, a_{21} < 0$

Observation:

- long-time behaviour depends on size of parameters
- steady-state solutions exist
- some scenarios are physically incorrect! (negative population size)

A Non-Linear Model

- similar to Verhulst's logistic growth model
- additional growth term proportional to other species
- leads to system of differential equations:

$$\begin{aligned}\dot{p}(t) &= (b_1 + a_{11}p(t) + a_{12}q(t))p(t) \\ \dot{q}(t) &= (b_2 + a_{21}p(t) + a_{22}q(t))q(t)\end{aligned}$$

- typically:
 - $b_1 > 0, b_2 > 0$ (growth term)
 - $a_{11} < 0, a_{22} < 0$ (saturation)
 - $a_{12}, a_{21}?$

The Non-Linear Competition Model

- two species sharing a common natural habitat
- competition: $a_{12} < 0, a_{21} < 0$

Possible Scenarios:

- steady-state
- one species dies out (extinction)
- no obvious nonsense

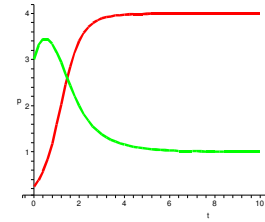
Competition – Steady State

- system of differential equations:

$$\dot{p}(t) = \left(\frac{5}{2} + \frac{\sqrt{3}}{24} - \frac{5}{8}p(t) - \frac{\sqrt{3}}{24}q(t) \right) p(t)$$

$$\dot{q}(t) = \left(\frac{7}{8} + \frac{3\sqrt{3}}{2} - \frac{3\sqrt{3}}{8}p(t) - \frac{7}{8}q(t) \right) q(t)$$

- solution for $p_0 = \frac{1}{4}, q_0 = 3$:



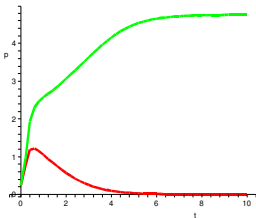
Competition – Extinction

- system of differential equations:

$$\dot{p}(t) = \left(\frac{71}{8} - \frac{23}{12}p(t) - \frac{25}{12}q(t) \right) p(t)$$

$$\dot{q}(t) = \left(\frac{73}{8} - \frac{25}{12}p(t) - \frac{23}{12}q(t) \right) q(t)$$

- solution for $p_0 = \frac{1}{4}, q_0 = \frac{1}{4}$:



Predator-Prey

- two species: predator p and prey q
- predator eats prey: $a_{12} > 0$
- prey is eaten by predator: $a_{21} < 0$

Possible Scenarios:

- stable oscillations
- one species dies out (what happens with the other, then?)
- Classical scenario: predator-prey equations by Lotka (1925) and Volterra (1926)

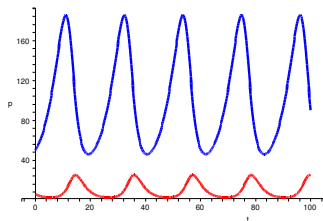
Predator-Prey by Lotka & Volterra

- system of differential equations:

$$\dot{p}(t) = \left(-\frac{1}{2} + \frac{1}{200}q(t) \right) p(t)$$

$$\dot{q}(t) = \left(\frac{1}{5} - \frac{1}{50}p(t) \right) q(t)$$

- solution for $p_0 = 6, q_0 = 50$:



Part IV

Analysis of ODE Models – Two Species

Critical Points in 2D

Example: Arms Race

- system of differential equations
- equilibrium: $\dot{p} = 0, \dot{q} = 0$

$$\dot{p}(t) = b_1 + a_{11}p(t) + a_{12}q(t) = 0$$

$$\dot{q}(t) = b_2 + a_{21}p(t) + a_{22}q(t) = 0$$

- solution of a linear system of equations:

$$a_{11}p(t) + a_{12}q(t) = -b_1$$

$$a_{21}p(t) + a_{22}q(t) = -b_2$$

- in most cases one critical point
- critical line, if system matrix is singular

- example: 2D system of differential equations:

$$\dot{p}(t) = b_1 + a_{11}p(t) + a_{12}q(t)$$

$$\dot{q}(t) = b_2 + a_{21}p(t) + a_{22}q(t)$$

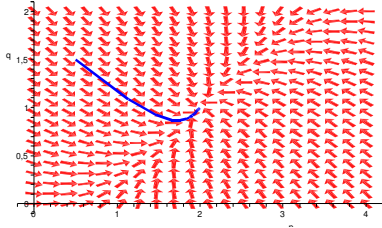
- natural extension: 3D plot: t vs. p vs. q
- 1D direction field for p vs. t or q vs. t not sufficient: what values to chose for q (or p resp.)?
- but: stationary problem \Rightarrow independent of t
- thus: plot directions depending on p and q

2D Direction Field – Arms Race

- system of differential equations:

$$\begin{aligned} \dot{p}(t) &= \frac{3}{2} - p(t) + \frac{1}{2}q(t) \\ \dot{q}(t) &= 0 + \frac{1}{2}p(t) - q(t) \end{aligned}$$

- direction field – with critical point at (2, 1):

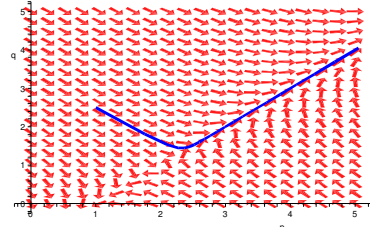


Arms Race – unlimited growth

- system of differential equations:

$$\begin{aligned} \dot{p}(t) &= \frac{1}{2} - \frac{3}{4}p(t) + q(t) \\ \dot{q}(t) &= -\frac{5}{4} + p(t) - \frac{3}{4}q(t) \end{aligned}$$

- direction field – with critical point at (2, 1):

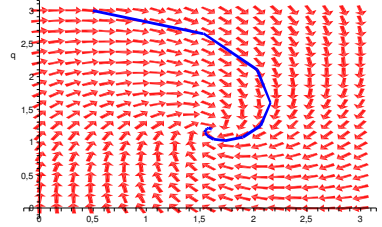


Arms race – the peaceful neighbour

- system of differential equations:

$$\begin{aligned} \dot{p}(t) &= 0 - \frac{3}{4}p(t) + q(t) \\ \dot{q}(t) &= \frac{5}{2} - p(t) - \frac{3}{4}q(t) \end{aligned}$$

- direction field – with critical point at $(\frac{6}{5}, \frac{6}{5})$:

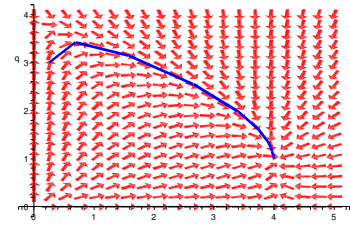


Nonlinear System – Competition

- system of differential equations:

$$\begin{aligned} \dot{p}(t) &= \left(\frac{5}{2} + \frac{\sqrt{3}}{24} - \frac{5}{8}p(t) - \frac{\sqrt{3}}{24}q(t) \right) p(t) \\ \dot{q}(t) &= \left(\frac{7}{8} + \frac{3\sqrt{3}}{2} - \frac{3\sqrt{3}}{8}p(t) - \frac{7}{8}q(t) \right) q(t) \end{aligned}$$

- direction field – critical points at (4, 1), ...:

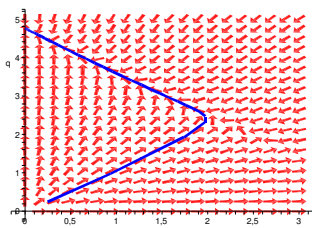


Nonlinear System – Extinction

- system of differential equations:

$$\begin{aligned} \dot{p}(t) &= \left(\frac{71}{8} - \frac{23}{12}p(t) - \frac{25}{12}q(t) \right) p(t) \\ \dot{q}(t) &= \left(\frac{73}{8} - \frac{25}{12}p(t) - \frac{23}{12}q(t) \right) q(t) \end{aligned}$$

- critical points at (0, 4.76...), (4.63..., 0), ...:

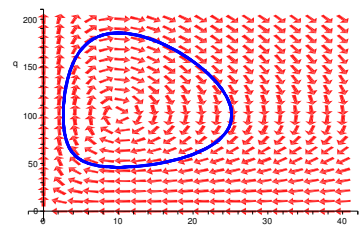


Lotka & Volterra

- system of differential equations:

$$\begin{aligned} \dot{p}(t) &= \left(-\frac{1}{2} + \frac{1}{200}q(t) \right) p(t) \\ \dot{q}(t) &= \left(\frac{1}{5} - \frac{1}{50}p(t) \right) q(t) \end{aligned}$$

- direction field – with critical point at (10, 100):



2D Critical Points – Summary

Different types of critical points in 2D:

- attractive/stable equilibrium (arms race – steady state)
- unstable equilibrium
- saddle point (arms race – unlimited growth)
- attractive “spiral point” (“peaceful neighbour”)
- unstable “spiral point”
- centre of “rotation” (Lotka-Volterra)

⇒ How to discriminate between these types?

Homogeneous Systems of ODE

Homogeneous System in matrix-vector-notation:

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x}$$

- $\mathbf{x} : \mathbb{R} \rightarrow \mathbb{R}^n, \mathbf{A} \in \mathbb{R}^{n \times n}$
- example: $\mathbf{x}(t) = (p(t), q(t))$

Solutions:

- let \mathbf{x}_λ be an eigenvector: $\mathbf{A}\mathbf{x}_\lambda = \lambda\mathbf{x}_\lambda$
- then $\mathbf{x}_\lambda e^{\lambda t}$ is a solution:

$$\mathbf{A}\mathbf{x}_\lambda e^{\lambda t} = \lambda\mathbf{x}_\lambda e^{\lambda t} = \frac{d}{dt}(\mathbf{x}_\lambda e^{\lambda t}) \quad \text{q.e.d.}$$

Eigenvectors and Eigenvalues

Corollaries:

- the solutions of the homogeneous system $\dot{\mathbf{x}} = \mathbf{Ax}$ are linear combinations of the respective eigen-solutions:

$$\mathbf{x}_{\text{hom}}(t) = \sum_{\lambda} a_{\lambda} \mathbf{x}_{\lambda} e^{\lambda t}, \quad a_{\lambda} \in \mathbb{R}$$

- the solutions of the inhomogeneous system $\dot{\mathbf{x}} = \mathbf{Ax} + \mathbf{b}$ are

$$\mathbf{x}(t) = -\mathbf{A}^{-1} \mathbf{b} + \mathbf{x}_{\text{hom}}(t)$$

- observation: $\mathbf{x}_c = -\mathbf{A}^{-1} \mathbf{b}$ is a critical point!

Eigenvalues and Critical Points

- the ODE system $\dot{\mathbf{x}} = \mathbf{Ax} + \mathbf{b}$ is solved by

$$\mathbf{x}(t) = \mathbf{x}_c + \sum_{\lambda} a_{\lambda} \mathbf{x}_{\lambda} e^{\lambda t}$$

- \mathbf{x}_c attractive equilibrium,

$$\lim_{t \rightarrow \infty} \mathbf{x}(t) = \mathbf{x}_c,$$

only if $e^{\lambda t} \rightarrow 0$ for all eigenvalues λ

- $\lambda \in \mathbb{R} \Rightarrow \lambda < 0$
- $\lambda = \mu + i\nu \Rightarrow \mu < 0$ ($e^{i\nu t} = \cos \nu t + i \sin \nu t$)

Stability of Linear Systems

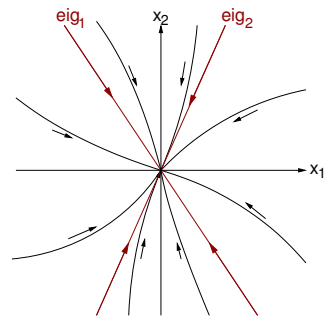
Overview:

eigenval. ($\lambda_j = \mu_j + i\nu_j$)	critical point	stability
real, all $\lambda < 0$	node	stable, attr.
real, all $\lambda > 0$	node	unstable
real, $\lambda_k > 0, \lambda_l < 0$	saddle point	unstable
complex, all $\mu < 0$	spiral point	stable, attr.
complex, all $\mu > 0$	spiral point	unstable
complex, all $\mu = 0$	centre	stable

Stability of 2D Systems

Real Eigenvalues:

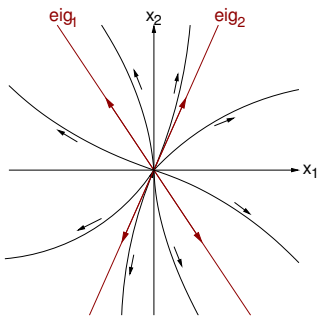
- $\lambda_1 < 0, \lambda_2 < 0$, attractive equilibrium



Stability of 2D Systems

Real Eigenvalues:

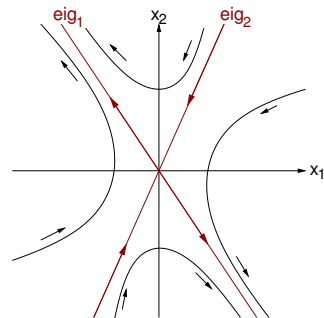
- $\lambda_1 > 0, \lambda_2 > 0$, unstable equilibrium



Stability of 2D Systems

Real Eigenvalues:

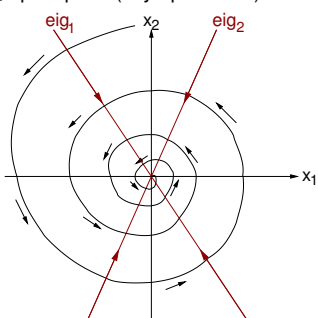
- $\lambda_1 > 0, \lambda_2 < 0$, saddle point



Stability of 2D Systems

Complex Eigenvalues:

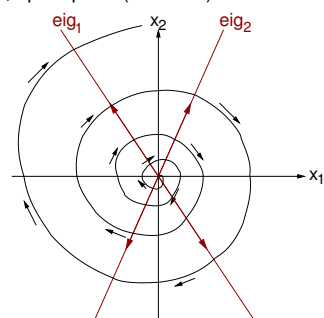
- $\mu_1 < 0, \mu_2 < 0$, spiral point (asympt. stable)



Stability of 2D Systems

Complex Eigenvalues:

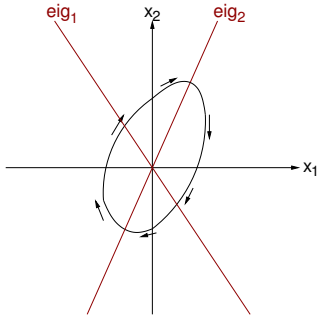
- $\mu_1 > 0, \mu_2 > 0$, spiral point (unstable)



Stability of 2D Systems

Complex Eigenvalues:

- $\mu_1 = \mu_2 = 0$, centre of oscillation



Stability of Non-Linear Systems

- 2D system of ODE:

$$\dot{\mathbf{x}}(t) = f(\mathbf{x}(t)),$$

$f : \mathbb{R}^n \rightarrow \mathbb{R}^n$ nonlinear

- critical point at \mathbf{x}_c : $f(\mathbf{x}_c) = 0$
- for analysis of critical points: linearization

$$\dot{\mathbf{x}}(t) = f(\mathbf{x}(t)) \approx \underbrace{f(\mathbf{x}_c)}_{=0} + \mathbf{J}_f(\mathbf{x}_c)(\mathbf{x}(t) - \mathbf{x}_c)$$

- examine eigenvalues of $\mathbf{J}_f(\mathbf{x}_c)$