

# **Summer school on collective behaviour**

Differential equations models

# Dynamical systems governed by ordinary differential equations

1. Geometric view
2. Linear stability analysis
3. Example

# Evolution of a system described by a set of Ordinary Differential Equations (O.D.E.)

$$\frac{dX_i}{dt} = F_i (X_1, \dots, X_n, \lambda)$$

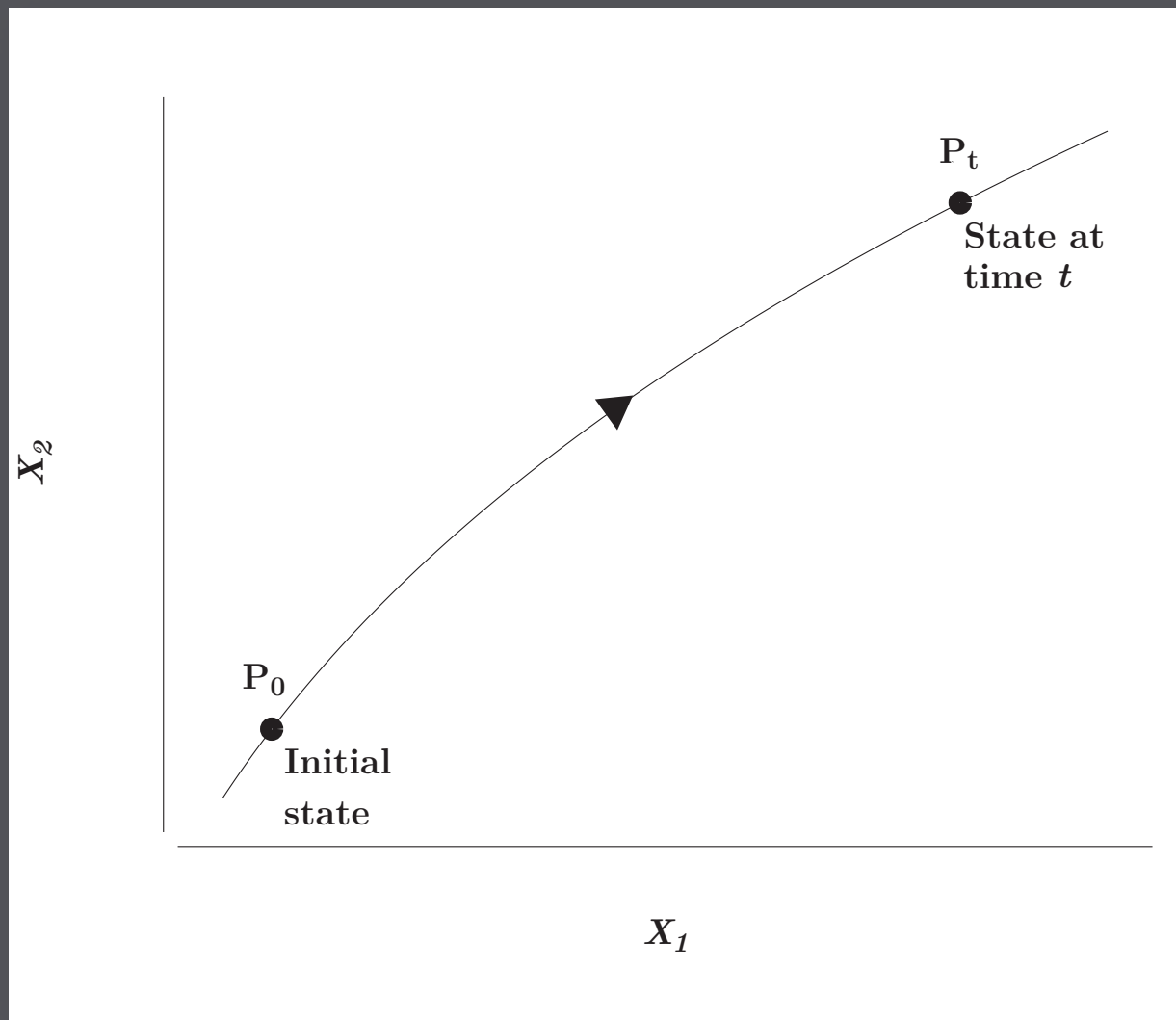
where

- $X_i$  are the variables describing the system (population densities, chemical concentrations, etc.)
- $\lambda$  are parameters (rate constants, birth rates, etc.)

**In presence of nonlinearities, no analytic solutions available !**

# 1. Geometric view

**Phase space** : embedding the evolution of the system into the  $n$ -dimensional space spanned by the full set of variables



## **phase space trajectory.**

The set of all phase space trajectories will provide all possible behaviors of our system.

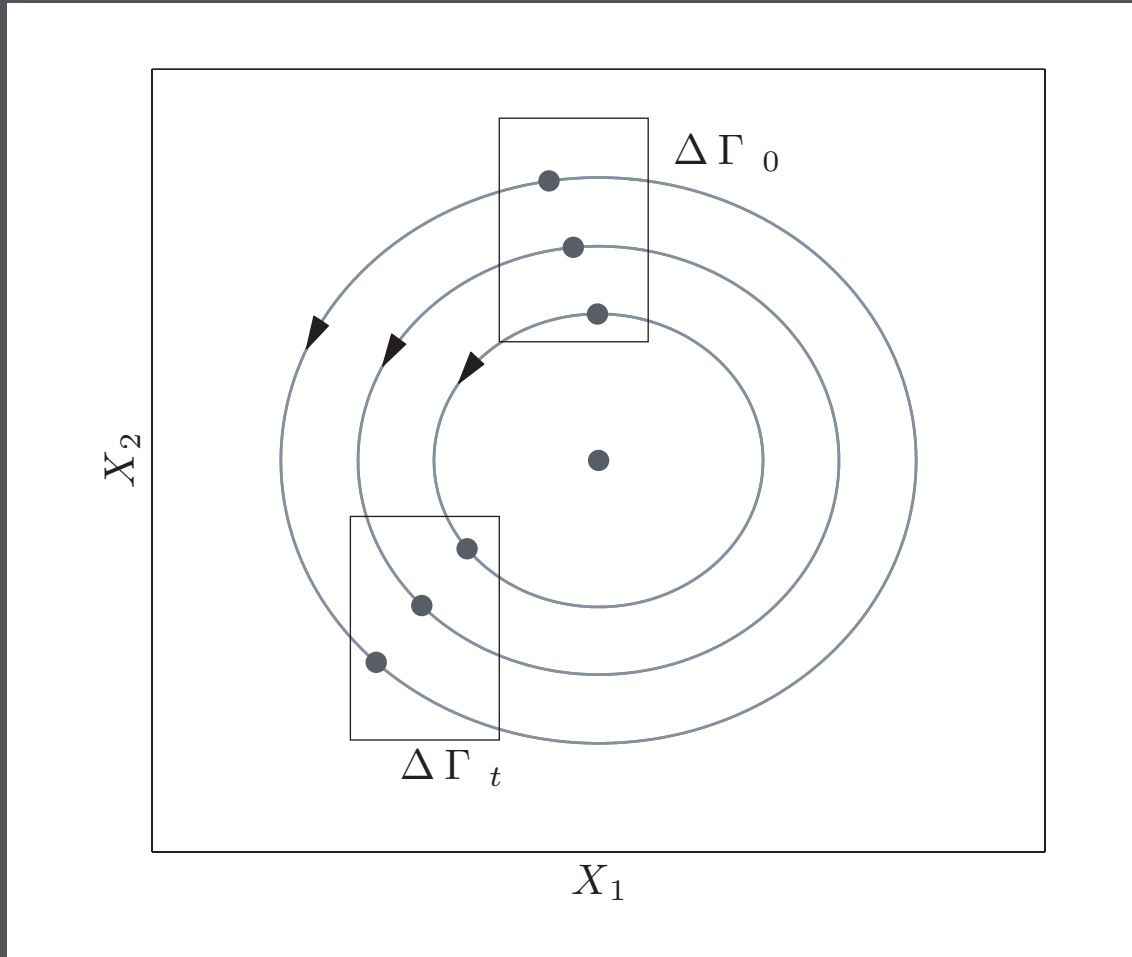
## **Invariant sets**

Objects in phase space mapped onto themselves during the time evolution, e.g.

- i. **fixed points** (describing the stationary states that can be reached by the system).
- ii. **closed curves** (describing a periodic behavior).

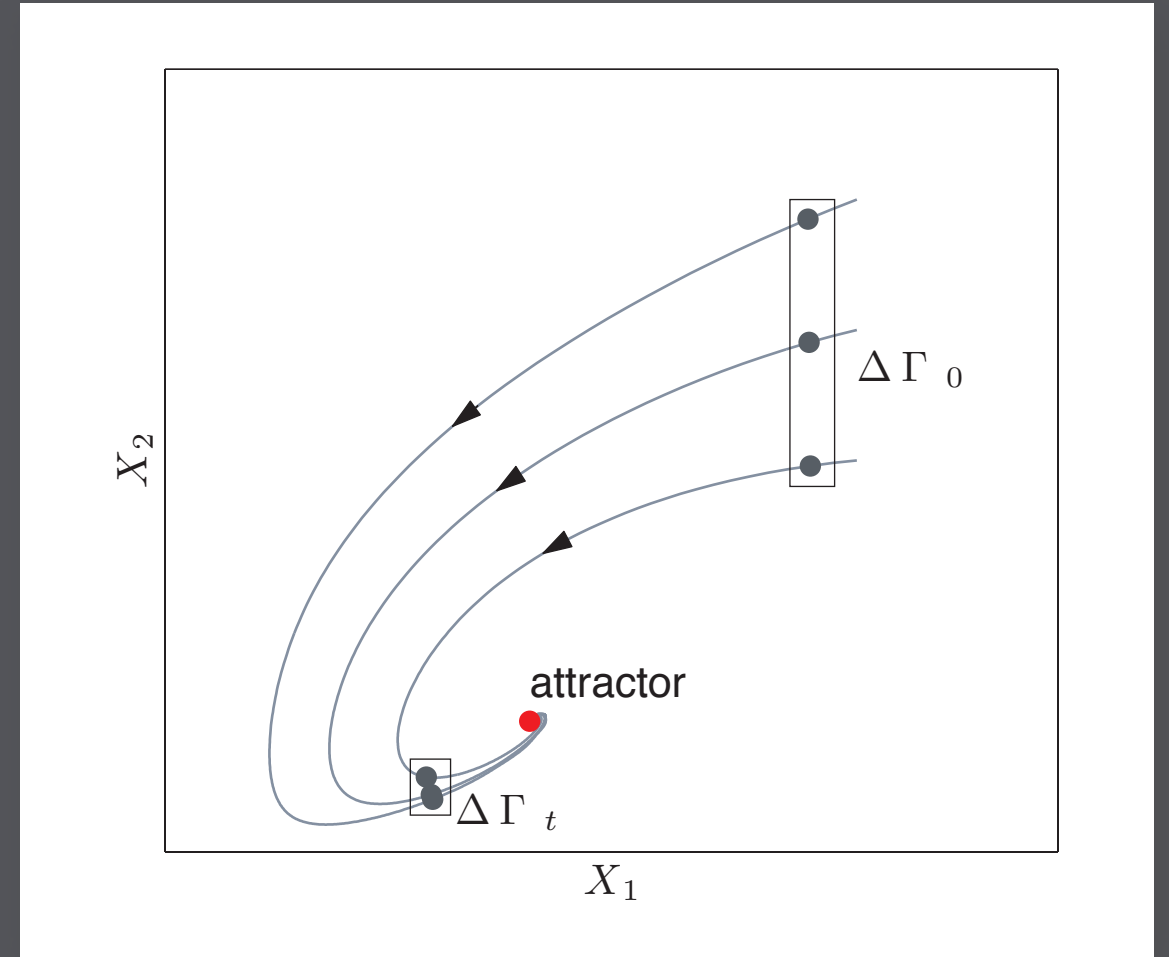
More complex invariant sets include tori and fractals, encountered in systems exhibiting quasi-periodicity or chaos.

## Conservative and dissipative systems



**Conservative system**

$$|\Delta\Gamma_0| = |\Delta\Gamma_t|$$



**Dissipative system**

$$|\Delta\Gamma_t| < |\Delta\Gamma_0|$$

**attractor** : ending up on a lower dimensionality

## 2. Stability

Response to a perturbation removing the system from an initial reference set,  $s$

$$X_i(t) = X_{i,s} + \delta x_i(t)$$

- **Stable system** : system remains in a neighborhood of the reference state,  $s$ .
- **Asymptotic stability** :  $\delta x_i(t) \rightarrow 0$  as  $t \rightarrow \infty$ .

Evolution of dynamical systems viewed as a problem of loss of stability of the reference state(s) (e.g., the fixed points) and the emergence of more intricate attractors.

# Linear stability analysis

$$\frac{dX_i}{dt} = F_i(\{X_j\}, \lambda) \quad j = 1, \dots, n$$

- Search for reference state, usually among the steady states

$$F_i(\{X_{j,s}\}, \lambda) = 0$$

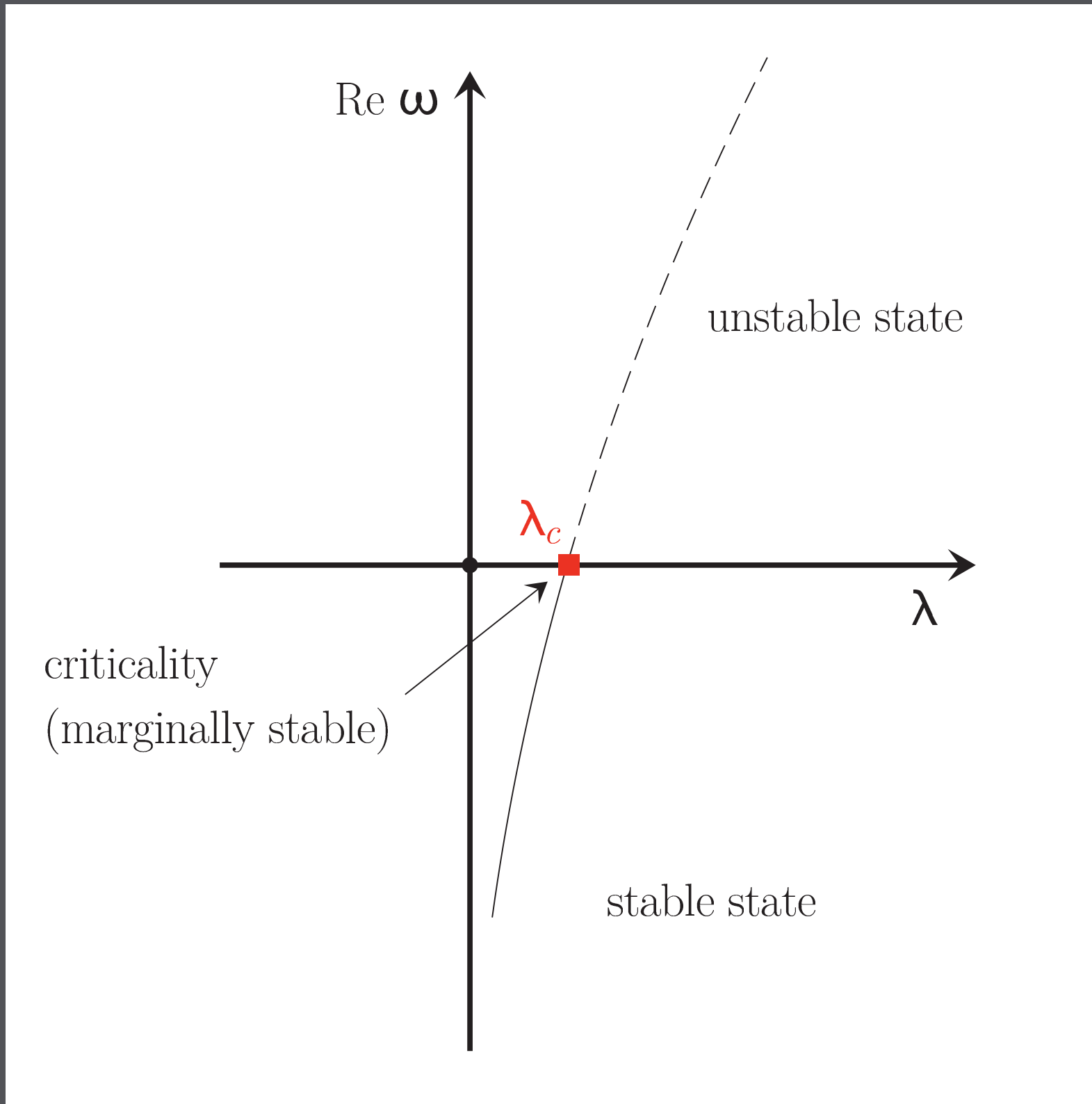
- Linearize around  $\{X_{j,s}\}$

$$X_j = X_{j,s} + \delta x_j \quad \frac{d\delta x_i}{dt} = \sum_j \left( \frac{\partial F_i}{\partial X_j} \right)_s \delta x_j \quad \begin{array}{l} \text{Solution in the form} \\ \delta x_i = u_i e^{\omega_\alpha t} \end{array}$$

- Determine the eigenvalues  $\omega_\alpha$  ( $\alpha = 1, \dots, n$ ) of the operator (Jacobian matrix)

$$J_{ij} = \left( \frac{\partial F_i}{\partial X_j} \right)_s$$

as roots of the characteristic equation  $\det \left| \left( \frac{\partial F_i}{\partial X_j} \right)_s - \omega \delta_{ij} \right| = 0$



**Critical value of the parameter at which the eigenvalue changes sign, i.e., the reference state changes stability.**



### 3. Example : Population biology

$$\frac{dX}{dt} = \text{birth} - \underbrace{\text{death}}_{k_2 X} \quad X : \text{ population density}$$

1. **Malthus view (18th century)** : birth rate,  $k_1 X$

But, this leads to **explosion** if  $k_1 > k_2$   
or to **extinction** if  $k_1 < k_2$

2. **Verhulst view (19th century)** : regulated birth rate,  $(k_1 - bX)X$

$$\frac{dX}{dt} = k_1 X - bX^2 - k_2 X$$

$$= \dots$$

$$= kX \left( 1 - \frac{X}{N} \right)$$

where  $k = k_1 - k_2$  and  $N = k/b$

stationary state :

$$X_{s_1} = 0$$

$$X_{s_2} = N = \frac{k}{b}$$

Stability :

$$X = X_s + \delta x$$

$$\frac{d\delta x}{dt} = \underbrace{\left( k - \frac{2kX_s}{N} \right)}_{(\partial F / \partial X)_s \equiv \omega} \delta x$$

$$\Rightarrow X_{s_1} = 0 \rightarrow \omega = k$$

$$X_{s_2} = N \rightarrow \omega = -k$$

bifurcation diagram :

