

Ordinary Differential Equation (ODE) models: the focus of this course

One independent variable, usually time:

- $\frac{dx}{dt} = f(x, t)$ the rate of change of x is a function of x and time
one state variable: one dimensional system
- $\frac{dx}{dt} = f(x, y, t)$
 $\frac{dy}{dt} = g(x, y, t)$ coupled ODEs, two state variables: two dimensional system

Partial differential equations involve two or more independent variables (such as time and space), e.g.

$$\frac{\partial y}{\partial t} = f(y, t, x) + D \frac{\partial^2 y}{\partial x^2}$$

where x represents a spatial co-ordinate. The second term on the right hand side of this **reaction-diffusion equation** depicts diffusion of y in space. PDEs are the topic of BMA 774

A **first order ODE** only involves first derivatives. (Remember the alternative notation for time derivatives: $dx/dt = \dot{x}$)

Second and **higher order ODEs** involve higher order derivatives.

Second order ODEs commonly arise in physics because acceleration is a second derivative. Classic examples:

- $\frac{d^2x}{dt^2} = -\omega^2 x$ describes the motion of a mass on a (perfect) spring: the simple harmonic oscillator.
- $\frac{d^2x}{dt^2} = -\frac{g}{L} \sin x$ describes the simple pendulum (x is the angle to the vertical)
(Since $\sin x \approx x$ when x is small, the simple harmonic oscillator provides a good approximation to the pendulum model when the oscillations have small amplitude.)

We focus on Systems of First Order Ordinary Differential Equations (coupled 1st order ODEs)

- These provide a natural description of many biological systems
- **We can always reduce an ODE of any order to a set of first order ODEs.**

Example: $\frac{d^2x}{dt^2} = -\omega^2 x$

We use a trick: write $y = \frac{dx}{dt}$, because this gives $\frac{d^2x}{dt^2} = \frac{d}{dt} \left(\frac{dx}{dt} \right) = \frac{dy}{dt}$. This lets us rewrite the 2nd order derivative as a first order derivative, and hence our 2nd order ODE as a coupled pair of 1st order ODEs (i.e. a two dimensional system):

$$\begin{aligned}\frac{dx}{dt} &= y \\ \frac{dy}{dt} &= -\omega^2 x\end{aligned}$$

This approach can be extended to reduce an n th order ODE to n coupled 1st order ODEs.

General form of n dimensional first order model

$$\begin{aligned}\frac{dx_1}{dt} &= f_1(x_1, x_2, \dots, x_n, t) \\ \frac{dx_2}{dt} &= f_2(x_1, x_2, \dots, x_n, t) \\ &\vdots \\ \frac{dx_n}{dt} &= f_n(x_1, x_2, \dots, x_n, t)\end{aligned}$$

Model is **linear** if the RHS are linear combinations of the x_i , i.e. are of the form $f_i(x_1, x_2, \dots, x_n, t) = \alpha_{i1}x_1 + \dots + \alpha_{in}x_n + \beta_i$.

This means the x_i only appear raised to the first power and there are no products of two or more x_i .

NOTE: The α and β can be functions of time **but not of x** .

Linear: $\frac{dx}{dt} = y$ Simple harmonic oscillator
 $\frac{dy}{dt} = -\omega^2 x$

Nonlinear: $\frac{dx}{dt} = -x^2$ $\frac{dx}{dt} = y$ $\frac{dx}{dt} = y$ Simple pendulum
 $\frac{dy}{dt} = -xy$ $\frac{dy}{dt} = -\frac{g}{L} \sin x$

Linear systems are easier to solve than nonlinear systems. (Many math courses will focus on linear ODEs for this reason.)

Reason: in some sense, we can break down a linear problem into smaller parts, each of which can be solved more easily, and can then put the parts back together to give a solution to the whole problem. This is not the case for nonlinear systems.

(Replacing $\sin x$ by x in the nonlinear pendulum model gives the linear SHO model. “Linearization” takes us from a model that is not easy to solve to a model that is easy to solve. This is an important approach that we will use quite often in 771.)

Autonomous and Non-Autonomous Systems

If the right hand side of the ODE does not depend **explicitly** on time, i.e. t does not appear on the RHS, then the system is **autonomous**. Otherwise, it is **non-autonomous**.

Some people use the term **forced** to describe a non-autonomous system. One common form of forcing involves a periodic function, e.g. a forced pendulum. In some contexts we talk about “seasonal forcing”.

(The right hand side of an autonomous ODE depends **implicitly** on time via the state variables.)

We will focus on autonomous systems. (It turns out that they have some nice properties...)

We can always convert a non-autonomous system to an autonomous one, at the cost of adding an extra dimension.

For example, $\dot{x} = f(x, t)$.

Write $\dot{z} = 1$, which gives $z = t + c$. c is some constant; set it equal to zero.

Then we can rewrite the non-autonomous ODE as the autonomous system:
 $\dot{x} = f(x, z)$
 $\dot{z} = 1$