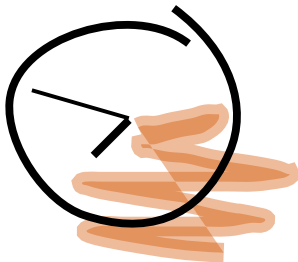


Numerical Algorithms (COMP 3371)

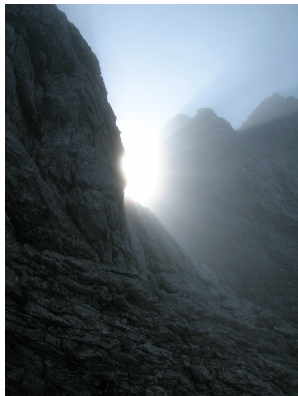
Session VI: ODEs and first glimpses of Euler

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Michaelmas term 2019

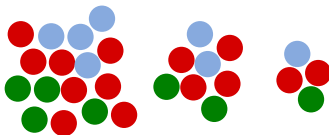


1. What is an *ill-posed* problem?
2. Why is an *ill-conditioned* problem?
Why are they problematic to solve?
3. Can an algorithm for an ill-conditioned problem be forward stable?
4. Write down the Taylor series.



Ordinary Differential Equations
Two model problems
Taylor and explicit Euler

Example: Radioactive decay



- ▶ Many problems in engineering, physics, social sciences are formulated in terms of functions that satisfy an equation with one or more derivatives.
- ▶ Example: Radioactive decay

$$\begin{aligned}\frac{\partial r(t)}{\partial t} &= -k \cdot r(t) \\ \partial_t r(t) &= -k \cdot r(t) \\ r'(t) &= -k \cdot r(t) \\ \dot{r}(t) &= -k \cdot r(t)\end{aligned}$$

with constant k and r being the amount of radioactive substance

- ▶ The four variants are different notations for the same thing
- ▶ Can you describe (informally) what the ODE means?

ODE: Let the function $f(t)$ depend on the time t . An ordinary differential equation (ODE) of order n is

$$F(t, f(t), f'(t), f''(t), f'''(t), \dots, f^{(n-1)}(t)) = f^{(n)}(t).$$

- ▶ Function $f(t)$ depends on its own derivatives.
- ▶ n above is the *order* of the ODE.
- ▶ $\underbrace{1 - t}_{F(t)} = f(t)$
Degenerated ODE (not really one)
- ▶ $\underbrace{1 - f(t)}_{F(t,f)} = f'(t)$
First order ODE
- ⇒ Most physical phenomena (first order) are described this way
- ⇒ Only one derivative (∂_t or ∂_x). More than one derivative would construct partial differential equations (PDEs)

Some ODE terminology

Linear: An ODE is linear iff the individual derivatives of f are isolated:

$$f'(t) = f^2(t) \quad \text{non-linear}$$

$$f''(t) = f'(t)f(t) + t \quad \text{non-linear}$$

$$f'(t) = f(t) \cdot t \quad \text{linear}$$

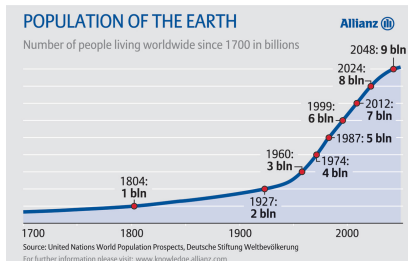
Autonomous: An ODE is autonomous if F does not directly depend on t (but only indirectly through $f(t)$):

$$f''(t) = f'(t)f(t) + t \quad \text{not autonomous}$$

$$f'(t) = f(t) \cdot t \quad \text{not autonomous}$$

$$f'(t) = f(t) + 2 \quad \text{autonomous}$$

Example: Model of Malthus



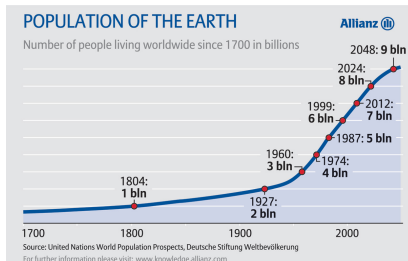
- ▶ Thomas Robert Malthus: *An Essay on the Principle of Population* (1798)
- ▶ Study one type of species (rabbits in Australia, e.g.)
- ▶ Malthus' assumptions:

1. The ratio of rabbits reproducing is constant: γ
2. The ratio of rabbits dying is constant: δ

$$\partial_t p(t) = (\gamma - \delta)p(t)$$

- ▶ We know the solution to this equation
- ▶ Is it a realistic model (same model was used also in economics sometimes)?

Example: Model of Malthus

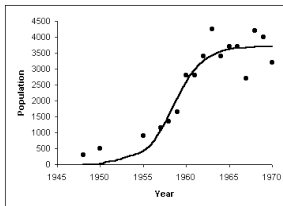


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- ▶ Model was realistic for population growth between 1700–1960!

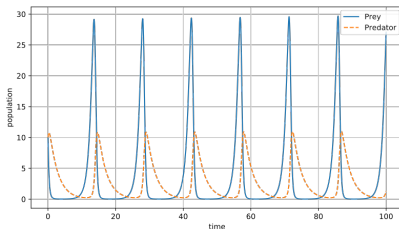
Example: Model of Verhulst



Barnacle Goose population (from Armson, Cockroft and Stone, 2000)

- Established by Verhulst in 19th century
- Extension: due to limited food
 1. the bigger the population the higher the death rates
 2. the bigger the population the lower the birth rates

$$\begin{aligned}\partial_t p(t) &= \gamma(t) - \delta(t) \\ \gamma(t) &= \gamma_0 - \gamma_1 p(t) \\ \delta(t) &= \delta_0 + \delta_1 p(t) \quad \text{or simply} \\ \partial_t p(t) &= b - ap(t)\end{aligned}$$



From Wikipedia.

- Our f (or p or whatever used) can be vector, i.e. $(f_1, f_2)(t)$
- Predator-prey model (Lotka-Volterra):

$$\partial_t p_1(t) = \alpha p_1(t) - \beta p_1(t)p_2(t)$$

$$\partial_t p_2(t) = \delta p_1(t)p_2(t) - \gamma p_2(t)$$

- p_1 is prey (rabbit)
 - The more rabbits, the faster the rabbit population growth (α)
 - The more rabbits and foxes, the more rabbits die (β)
- p_2 is predator (foxes) feeding on rabbits
 - The more rabbits and foxes, the higher the reproduction rate (δ)
 - The more foxes, the more foxes die (γ)

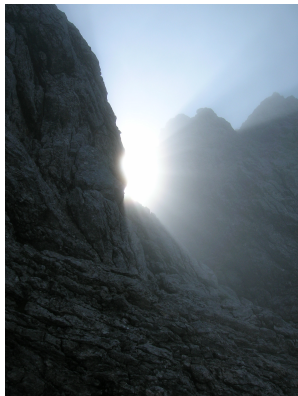
$$\begin{aligned} f''(t) + f(t) &= f'(t) \\ x_1(t) &:= f(t) & x_2(t) &:= f'(t) \end{aligned}$$

We obtain:

$$\begin{aligned} x_2'(t) + x_1(t) = x_2(t) &\Rightarrow x_2'(t) = x_2(t) - x_1(t) \\ x_1'(t) &= x_2(t) \end{aligned}$$

- ▶ For ODEs, we can mechanically replace higher-order terms by new (helper) functions
- ▶ We reduce the order of the ODE system but increase the entries of the vector
- ⇒ We don't have to study higher-order ODEs
- ⇒ This does not work (that way) for PDEs

- ▶ Content
 - ▶ Introduce term ODE at hands of radioactive decay
 - ▶ Define terminology around ODEs
 - ▶ Give a few examples of ODEs from population modelling
 - ▶ Discuss vector-valued ODEs
- ▶ Expected Learning Outcomes
 - ▶ Student knows fundamental ODE terminology
 - ▶ Student can explain population models (not to be memorised)
- ▶ Further reading
 - ▶ KHANAcademy: <https://www.khanacademy.org/math/differential-equations/first-order-differential-equations/modeling-with-differential-equations/v/modeling-population-with-simple-differential-equation>



Ordinary Differential Equations
Two model problems
Taylor and explicit Euler

Newton's law of gravity

- Gravity is a constant (9.8 here)
- Acceleration is the derivative of velocity with respect to time:

$$v'(t) = -9.8$$

- Position of particle is changed by velocity:

$$x'(t) = v(t)$$

- We obtain the second-order ODE:

$$x''(t) = -9.8$$

- First-order formulation:

$$\partial_t \begin{pmatrix} x \\ v \end{pmatrix} (t) = \begin{pmatrix} v \\ 0 \end{pmatrix} (t) + \begin{pmatrix} 0 \\ -9.8 \end{pmatrix}$$

⇒ This is something we can solve with A-level calculus

$$F = ma$$

- ▶ A particle is given by its position x , mass m and its velocity v
- ▶ x and v are vectors and they depend on the time t
- ▶ Definitions of x and v plus Newton's second law:

$$x'(t) = v(t)$$

$$v'(t) = F/m$$

$$F = \sum_i \frac{m \cdot m_i}{r^2} \left(\frac{x - x_i}{r} \right) \quad r \text{ depends on all the } x$$

- ▶ This is a system of two ordinary differential equations (ODEs), i.e. functions of type

$$\begin{pmatrix} x \\ v \end{pmatrix}'(t) = \begin{pmatrix} v(t) \\ \frac{1}{m} \sum_i \frac{m \cdot m_i}{r^2} \left(\frac{x(t) - x_i(t)}{r} \right) \end{pmatrix}$$

- ⇒ Not that easy to solve anymore
- ⇒ For N particles, we end up with $3 \cdot 2 \cdot N$ array entries (doubles)

- ▶ We know know where our coursework's equations come from, but we have to solve them
 - ▶ High school approach (symbolic):
 - ▶ Search for fundamental solution (integrate \Rightarrow obtain some magic constants)
 - ▶ Plug given values into solution
 - ▶ Closed expression represents *all* solutions
 - ▶ Our approach (numerical):
 - ▶ Write code that approximates solution $f(t)$
 - ▶ Take given initial values and plug them into code
 - ▶ Run the code
 - ▶ Numerical result represents *one* solution
- \Rightarrow To “solve” an ODE, we always need initial value (*Initial Value Problem (IVP)*)
Without proper/valid initial values, problem would be ill-posed

IVP in action

Radioactive decay—an analytical solution

$$\begin{aligned}\partial_t r(t) &= -k \cdot r(t) \\ \text{ansatz: } r(t) &= ae^{bt}\end{aligned}$$

Our ansatz here stems from experience/intuition. In general, it is not that trivial (maybe even impossible). Then, we need numerical approximation.

IVP in action

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$$\begin{aligned}\text{ansatz: } r(t) &= ae^{bt} \\ \Rightarrow \partial_t r(t) &= abe^{bt} \\ &= b \cdot r(t) \\ \Rightarrow b &= -k \\ \Rightarrow r(t) &= ae^{-kt}\end{aligned}$$

So there's an infinite number of solutions for all the a ?

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So there's an infinite number of solutions for all the a ?

No, because we have an initial amount of substance $r(t)$ that fixes a .



- ▶ Radioactive decay had been motivating example, but ...
⇒ abstract from application

$$y'(t) = \lambda y(t) \quad \text{with} \quad y(0) = y_0$$

- ▶ Perfect test equation
 - ▶ Simple
 - ▶ We know analytical solution

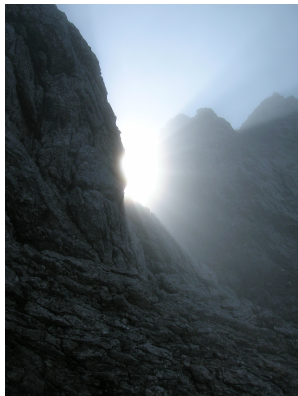
$$y(t) = y_0 e^{\lambda t}$$

- ⇒ easy to compare/measure errors
- ⇒ well-posed problem

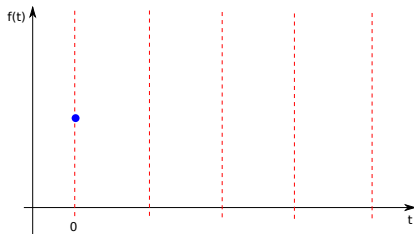
- ▶ Interesting behaviour (as we'll see)
- ▶ Called Dahlquist's test equation
- ▶ Condition number: $\kappa_{y_0} = \left| \frac{y_0 \cdot e^{\lambda t}}{y_0 e^{\lambda t}} \right|$
⇒ without analytic solution, computing κ would be nasty
⇒ well-conditioned problem

Concept of building block

- ▶ Content
 - ▶ Derive model problem of coursework from school physics
 - ▶ Introduce Dahlquist test equation
- ▶ Expected Learning Outcomes
 - ▶ Student knows Dahlquist test equation
 - ▶ Student can explain where coursework ODE comes from
 - ▶ Student can show properties of test equation (well-posedness, conditioning)

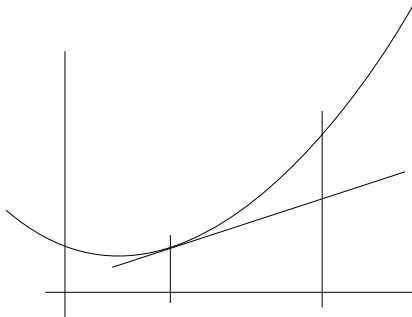


Ordinary Differential Equations
Two model problems
Taylor and explicit Euler



$$f'(f(t), t) = F(f, t) \quad \text{with } f(0) = f_0$$

- ▶ Problem with ODEs: $f(t)$ is implicitly given while we would like to have it explicitly
 - ▶ Time slicing: We cut the solution into snapshots (with fixed time step sizes Δt for the time being)
- ⇒ We are interested in the numerical solution f_h at fixed time steps
- ▶ We know $f(0) = f_h(0)$
 - ▶ We make certain assumptions/simplifications about the solution in-between two snapshots



- ▶ Assume $f_h(t)$ known
- ▶ Use Taylor:

$$f_h(t + \Delta t) = f(t) + \frac{\Delta t^1}{1!} f'(t) + \frac{\Delta t^2}{2!} f''(t) + \frac{\Delta t^3}{3!} f'''(t) + \dots$$

- ▶ Cut after second term
- ▶ Insert ODE: $f'(t) = F(f(t), t)$
- ▶ Walk for Δt along derivative

$$f(t + \Delta t) = f(t) + \Delta t \cdot f'(t)$$

- ▶ We cut the Taylor series after the first term and end up with *explicit Euler*
- ▶ The source code realisation of this scheme is a simple for loop
- ▶ Explicit = we can step through time without running into linear equation systems
- ▶ It is a *single step method*, as a new solution only needs the previous one (Δt is often also denoted as h or dt)
- ▶ It is a *finite difference* method, as we use the difference quotient instead of the derivative:

$$f(t + \Delta t) = f(t) + \Delta t \cdot f'(t) \Leftrightarrow f'(t) = \frac{f(t + \Delta t) - f(t)}{\Delta t}$$

```
double t    = 0.0;
double f    = 4.0;
double dt   = 0.001;

while (t < 2.0) {                                // simulate until t=2.0
    t = t + dt;
    f = f + dt * F(f, t);                        // you have to implement F
    std::cout << "t=" << t << ", f=" << f << std::endl;
}
```

Concept of building block

- ▶ Content
 - ▶ Introduce idea of explicit Euler at hands of Taylor expansion
 - ▶ Motivate Euler with “walk along derivative” sketch
 - ▶ Show pseudo code
- ▶ Expected Learning Outcomes
 - ▶ Student *knows* the definition of explicit Euler
 - ▶ Student can *sketch* idea how to derive explicit Euler

Concepts discussed:

- ▶ We have defined condition number
- ▶ We have found a formal derivation of the explicit Euler
- ▶ We have studied some experiments we'd done before

Next:

- ▶ We study Euler's error in detail
- ▶ We find that there are further notions of unstable

Preparation:

- ▶ Revise the section on radioactive decay as ODE example and reiterate how to derive the analytic solution
- ▶ Solve the problem **Explicit Euler for some test ODEs** from the worksheet