Differential equation modelling

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Outline

Differential equation modelling

Phase Plane analysis

1st order: Euler

$$\frac{\mathrm{d}y}{\mathrm{d}x} = f(x,y)$$

For a fixed time-step h we Taylor expand to first order.

$$y(x+h) = y(x) + h\frac{dy}{dx} + h.o.t.$$

$$y(x+h) = y(x) + hf(x,y)$$

Making time-steps too small could cause accumulation of round-off error (Strogatz).

2nd order: midpoint method

Taylor expand to 2nd order; avoid calculating f'() by chain rule, and equate coefficients.

$$k_0 = hf(x, y(x))$$

$$k_1 = hf(x + h/2, y(x) + \frac{1}{2}k_0)$$

$$y(x + h) = y(x) + k_1$$

- More evaluations, but better accuracy than Euler.
- Make a guess at an initial step
- Or take principled approach to get family of solvers for different coefficients (4 unknown with 3 equations)

4th order Runge-Kutta formula (fixed step size)

$$k_0 = hf(x, y(x))$$

$$k_1 = hf(x + h/2, y(x) + \frac{1}{2}k_0)$$

$$k_2 = hf(x + h/2, y(x) + \frac{1}{2}k_1)$$

$$k_3 = hf(x + h, y(x) + k_2)$$

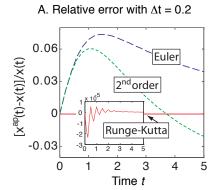
$$y(x + h) = y(x) + \frac{k_0}{6} + \frac{k_1}{3} + \frac{k_2}{3} + \frac{k_3}{6}$$

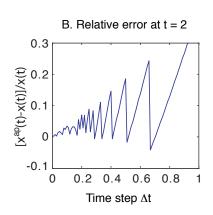
- f() may be expensive to compute.
- good trade off: computations performed vs accurary

Comparing solutions on problem with known solution

$$\frac{\mathrm{d}x}{\mathrm{d}t} = t - x + 1 \qquad x(0) = 1$$

$$\Rightarrow x(t) = t + e^{-t}$$





http://web.cs.dal.ca/~tt/fundamentals/figures/Appendices/

Adaptive step size solvers

- Compare accurary of one big jump (x+2h), with two jumps of size h.
- Cost: 11 f() evaluations, compared with 8 f() evaluations for two small jumps; e.g. figure 17.2.1 of NR.
- Adjust h to ensure that $\Delta \equiv y_2 y_1$ within tolerance.
- popular solver: ode45
- ▶ More efficient step-size methods now available.
- tolerance on error. rk45 popular.
- Step-size increases when solution smooth.
- Stiff systems are a problem. (One component changing rapidly comparing to others.)
- ▶ Use industrial-strength solvers, e.g. LSODA.

Framework for integration

- Code up function that returns derivatives at any time, given params.
- ► See DMB guide for examples.

Example: van der Pol oscillator

IVP (initial value problems) e.g. van der Pol oscillator:

$$y'' - \mu(1 - y^2)y' + y = 0$$

Reduce to 1st order system ("normal form"):

$$x = y'$$

 $x' - \mu(1 - y^2)x + y = 0$

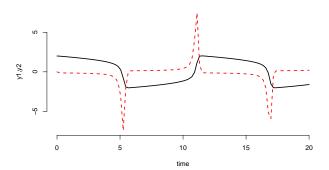
If we regard our state vector as $\begin{pmatrix} y \\ x \end{pmatrix}$ then we have:

$$\begin{pmatrix} y' \\ x' \end{pmatrix} = \begin{pmatrix} x \\ \mu(1-y^2)x - y \end{pmatrix}$$

Calling the solver

```
library(deSolve)
vanderpol = function(t, y, parms) {
  ## T = current time
  ## Y = current state vector
  ## PARMS = vector of parameters for system.
  mu = parms[1]
  derivs = c(y[2], mu*(1-y[1]^2)*y[2] - y[1])
  ## Return
  list(derivs)
init.cond = c(2, 0); parms = c(mu=5.0)
times = seq(from=0, to=20, length=100)
out = lsoda(init.cond, times, vanderpol, parms)
pdf(file='vanderpol.pdf', width=8, height=5); par(bty='n')
matplot(out[,1], out[,-1], type='l',xlab='time',
        ylab='y1,y2',lwd=2)
dev.off()
```

Output



Limitations

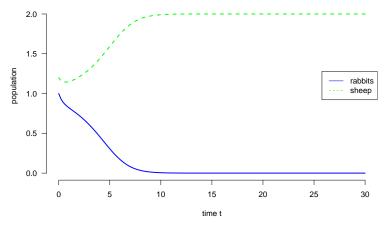
- How to add spikes into framework, if just returning derivative?
- or more generally, sudden changes to variables?
- how to debug?

Example: rabbits vs sheep (Strogatz, p155)

$$\frac{dr}{dt} = r(3 - r - 2s)$$
$$\frac{ds}{dt} = s(2 - r - s)$$

- 1. Compute trajectory over time
- 2. Plot phase planes
- 3. Nullclines
- 4. Examining stability

1. Computing any trajectory over time, as before.

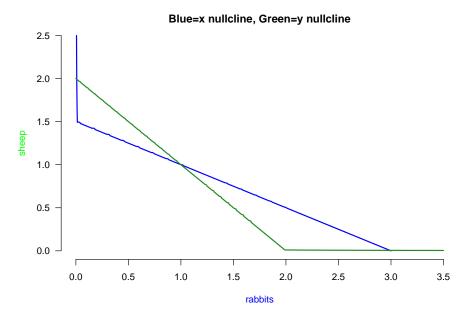


i.e. use numerical integration

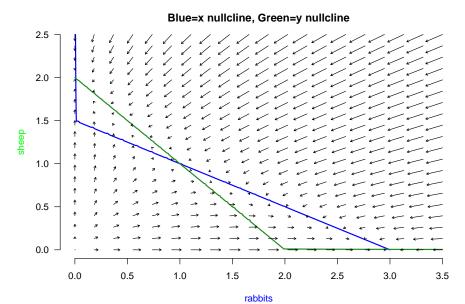
Finding nullclines and steady-states

- A nullcline for a variable x is where its DE $\frac{dx}{dt}$ is zero.
- ► For a two-variable system *x* and *y*, steady-states are found where the two nullclines intersect.

Poor-man's phase space: check either side of nullclines.



2. Phase-plane and 3. Nullclines



Steady-states: intersection of nullclines

Careful: not all nullclines are drawn on the previous page. This is an implementation issue with *contour()* for finding zeros.

- ► Find the steady-states
 - 1.
 - 2.
 - 3.
 - 4.

Classification of steady-states

Why take the Jacobian around the steady-state?

$$J(r,s) = \begin{pmatrix} 3-2r-2s & -2r \\ -s & 2-r-2s \end{pmatrix}$$

Recall: for a 2x2 matrix:

- ► Tr(J) = sum of eigenvalues
- Det(J) = product of eigenvalues.

Starting points

- deSolve package
- phase planes and nullclines (DMBpplane.r from DMB site, modified from Daniel Kaplan)
- phaseR package (by Michael Grayling, MGM)
- integrate() quadrature
- ► D() symbolic differentiation
- optimize() (1d) and optim() (n-d)
- Steven Strogatz. Nonlinear dynamics and chaos.
- ▶ NR: William Press et al. Numerical Recipes in C/C++
- ► Julia:

https://github.com/JuliaDiffEq/DifferentialEquations.jl