Numerical Integration Methods

Fish 559; Lecture 2



Some Preliminaries-I

- Numerical methods are numerical and hence subject to various types of errors:
 - Roundoff error: The consequences of a computer representing a real number using finite accuracy. Note:
 - roundoff errors accumulate; and
 - subtraction of two numbers that are nearly identical may lead to very large roundoff errors.
 - Truncation error: the difference between the true answer and the value from the numerical analysis.
 - e.g. approximation of a Taylor series which is an infinite series but we only use first two terms. E.g. delta method which uses the first term to estimate variance.



Some Preliminaries-II

Truncation error:

- There are often error bounds for methods (though we will not consider these in this course).
- Truncation error can be reduced by increasing the number of steps / reducing tolerances or step sizes, but this can be costly in terms of computation time and, eventually, roundoff error will begin to impact the results.
- E.g. setting low tolerance in TMB



Numerical Integration (or numerical quadrature)

The objective of numerical integration is to solve the problem:
b

$$I = \int_{a}^{b} f(x) \, dx$$

which is equivalent to finding y(b) subject to:

$$\frac{dy}{dx} = f(x); \ y(a) = 0$$

- We will assume that the function f cannot be integrated analytically!
- Many integrals are not analytic



Numerical Integration (why do this)

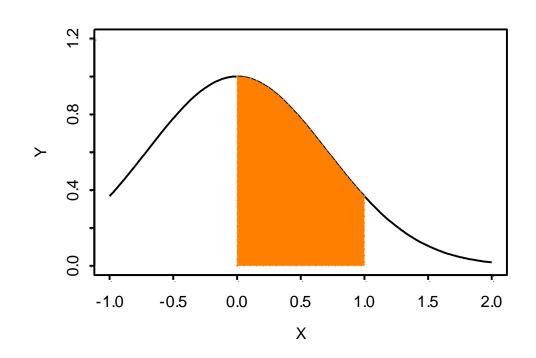
- The density of larvae is given by the function f (as a function of distance along a river); how many larvae are there between points a and b along the river?
- Let f be the function that determines the rate of change in the number of animals of species i (which depends on the number of animals of species i). If there are 1,000 animals of species i at time t_0 , how many are there at time t_0+1 .



Numerical Integration (An example problem)

We wish to evaluate the integral:

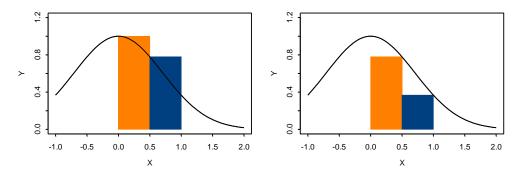
$$I = \int_{0}^{1} e^{-x^2} dx$$





Numerical Integration (Single-step approaches-I)

 We could approximate the integral by its height at its first point, at its last point, or at the middle.



 Approximating the integral by the average of the first two approaches is the "Trapezoidal rule" (only for this particular example).



Numerical Integration (Single-step approaches-II)

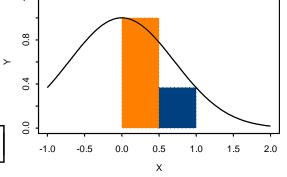
- We wish to integrate from a to b. We have formulae that get more accurate as more terms are added:
- Trapezoidal rule:

$$I \approx h\left[\frac{1}{2}f(a) + \frac{1}{2}f(b)\right]$$

Simpson's Rule:

$$I \approx h \left[\frac{1}{3} f(a) + \frac{4}{3} f(a+h) + \frac{1}{3} f(b) \right]^{\frac{2}{3}}$$

Boole's rule:



- $I \approx h \left[\frac{14}{45} f(a) + \frac{64}{45} f(a+h) + \frac{24}{45} f(a+2h) + \frac{64}{45} f(a+3h) + \frac{14}{45} f(b) \right]$
- h is the step size. It is (b-a) divided by the number of function calls less 1.
- There are many variants on this theme!



Numerical Integration (Single-step approaches-III)

Results:

- Trapezodinal rule: 0.68394
- Simpson's rule: 0.74718
- Boole's rule: 0.74683
- Exact: 0.74682!
- Tradeoff between truncation error and computation time.
- Acceptable level of accuracy depends on problem and rounding (accumulation of errors through multiple calculations)
- Most use Simpson's rule



Numerical Integration (Multi-step rules)

Now:
$$\int_{a}^{c} f(x)dx = \int_{a}^{b} f(x)dx + \int_{b}^{c} f(x)dx$$

• Therefore, in order to numerically integrate over the interval [a, c], we can apply one of the single-step rules multiple times and add the results, i.e. for Simpson's rule:

$$I \approx h \left[\frac{1}{3} f_1 + \frac{4}{3} f_2 + \frac{2}{3} f_3 + \dots + \frac{4}{3} f_{N-1} + \frac{1}{3} f_N \right]$$



Other Approaches (Gaussian quadrature-I)

Generalizing the previous approaches:

$$I = \sum_{i} w_{i} f(x_{i})$$

The previous methods have all been based on equal spacing along abscissa. We can improve our accuracy (at the expense of a bit more programming) by allowing freedom in the location of the absciccas.

$$I = \int_{a}^{b} f(x)dx = \int_{-1}^{1} f(x(t))x'(t)dt = \int_{-1}^{1} f(x(t))\frac{b-a}{2}dt = \frac{b-a}{2}\int_{-1}^{1} f(x(t))dt$$

$$I \approx \frac{b-a}{2}\sum_{i} w_{i} f(x(t_{i})) = \frac{b-a}{2}\sum_{i} w_{i} f''(t_{i})$$

 Tradeoff between truncation error and amount of programming effort to get it right



Other Approaches (Gaussian quadrature-II)

For two point Gaussian quadrature:

$$I \approx \frac{b-a}{2} \left[\tilde{f}(-0.57735026) + \tilde{f}(0.57735026) \right]$$

$$\int_{0}^{1} e^{-x^{2}} dx \approx \frac{1}{2} \left[e^{-(x+1)^{2}/4} + e^{-(1-x)^{2}/4} \right]; \quad x = 0.5773502691$$

- This lead to an approximation to the integral of 0.74965. The three point formula leads to 0.74673.
- Abramowitz and Stegun provide details on the absiccas and weights for equations of different orders (numbers of terms in the approximating equation).



The Error Function

Perhaps the most frequently required integral in fisheries is the "error function":

$$I = \int_{-\infty}^{a} e^{-x^2/2} dx$$

- It is not necessary to apply the methods of this lecture to compute this integral because very accurate approximations to it have been programmed and are available in software libraries (e.g. NORMDIST in EXCEL, cumd_norm in ADMB and pnorm in R and TMB).
- The lesson from this example, is that someone may have already solved your problem; before writing your own code check whether this is the case!



Multidimensional Integrals

- Multidimensional integrals are much more complicated than single-dimensional integrals. There are a few basic approaches to computing these integrals:
 - Split the problem into multiple single dimensional integrals.
 - Use Monte Carlo methods.
 - Use numerical methods (such as Laplace's method).
- The issue of computing multidimensional integrals using Monte Carlo methods will be discussed later.



Integration of ODEs

The generic problem is the study of a set of N coupled first order differential equations (e.g. EcoSim x = time, y = biomass of different components): vector of equations

$$\frac{dy_i(x)}{dx} = f_i(x, y_1, ..., y_N); \qquad i = 1, 2,, N$$

The specification of ODEs generally includes initial conditions, i.e.:

$$y_i(0) = Y_i;$$
 $i = 1, 2, ..., N$



A Familiar Set of ODEs (Ecosim)

Consumption of species *j* by species *i*

Non-predation mortality, fishing mortality and emigration

 The initial conditions are determined from Ecopath.

Euler's Method

• Euler's method updates the solution from x_n to $x_{n+1} = x_{n+h}$ where x is time using the formula:

$$y_{n+1} = y_n + hf(x_n, y_n)$$

 This method is simple, but can be very inaccurate (unless a very small step-size is used).



Other Ways to Solve ODEs

- Runge-Kutta methods.
- Richardson extrapolation.
- Predictor-corrector methods.

We will focus on the Runge-Kutta method.



Runge-Kutta Methods-I

- These tend to perform adequately for many problems (but more efficient algorithms do exist).
- The step size (h) can be monitored and modified automatically to improve performance.



Runge-Kutta Methods-II

 There are many versions of the Runge-Kutta method but the fourth-order version seems the most popular (evaluation of differential equations at different steps of x and y):

$$y_{n+1} = y_n + \frac{k_1}{6} + \frac{k_2}{3} + \frac{k_3}{3} + \frac{k_4}{6}$$

$$k_1 = hf(x_n, y_n)$$

$$k_2 = hf(x_n + \frac{h}{2}, y_n + \frac{k_1}{2})$$

$$k_3 = hf(x_n + \frac{h}{2}, y_n + \frac{k_2}{2})$$

$$k_4 = hf(x_n + h, y_n + k_3)$$

There are multidimensional versions of the RK algorithm.

An Example-I

Solve the following ODE using (a) Euler's method and (b) the Runge-Kutta fourthorder method:

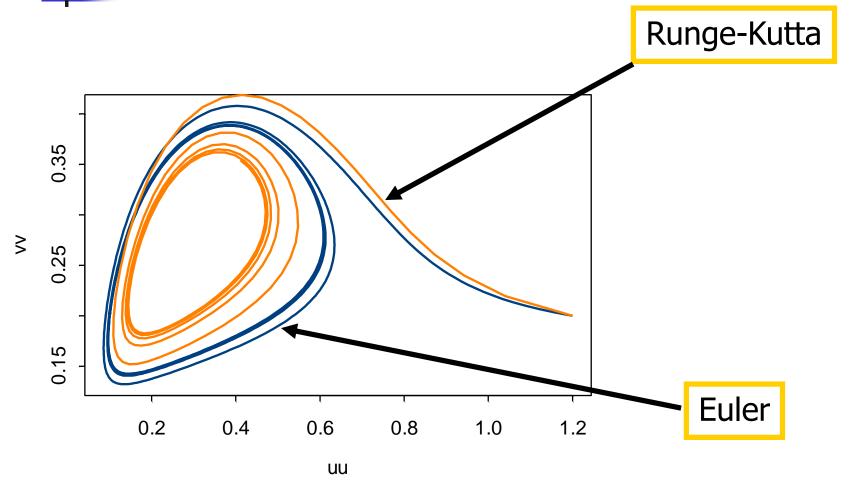
$$\frac{du}{dt} = u(1-u) - \frac{auv}{u+d}; \qquad \frac{dv}{dt} = bv(1-v/u)$$

where:

$$u(0) = 1.2$$
; $v(0) = 0.2$; $a = 1$; $b = 0.2$; $d = 0.1$



An Example -II





Moving to PDEs

- Partial Differential Equations (PDEs) are much more difficult to solve.
- Although algorithms do exist to solve these, the literature is vast and we will not discuss them in this course.



Numerical Instability-I

• We wish to solve the following initial value problem:

$$\frac{dy}{dx} = -2y + 1 \qquad y(0) = 1$$

We can apply Euler's method:

$$y_{n+1} = y_n + h f_n$$

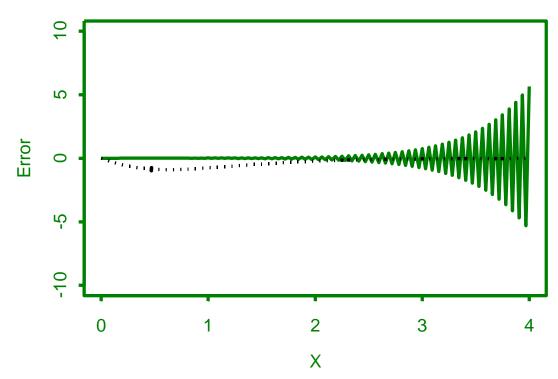
or Milne's method:

$$y_{n+1} = y_{n-1} + 2hf_n$$



Numerical Instability-II

Euler's method is stable but Milne's rule Becomes unstable after x=2!



In the case, we knew the correct solution, what would we do in reality?