

Mathematical Tripos Part IB: Lent 2019

Numerical Analysis – Lecture 9¹

Formally, $\mathbf{y}(t_{n+1}) = \mathbf{y}(t_n) + \int_{t_n}^{t_{n+1}} \mathbf{f}(t, \mathbf{y}(t)) dt$, and this can be ‘approximated’ by

$$\mathbf{y}_{n+1} = \mathbf{y}_n + h \sum_{l=1}^{\nu} b_l \mathbf{f}(t_n + c_l h, \mathbf{y}(t_n + c_l h)). \quad (4.11)$$

except that, of course, the vectors $\mathbf{y}(t_n + c_l h)$ are unknown! *Runge–Kutta methods* are a means of implementing (4.11) by replacing unknown values of \mathbf{y} by suitable linear combinations. The general form of a ν -stage explicit Runge–Kutta method (RK) is

$$\begin{aligned} \mathbf{k}_1 &= \mathbf{f}(t_n, \mathbf{y}_n), \\ \mathbf{k}_2 &= \mathbf{f}(t_n + c_2 h, \mathbf{y}_n + h c_2 \mathbf{k}_1), \\ \mathbf{k}_3 &= \mathbf{f}(t_n + c_3 h, \mathbf{y}_n + h(a_{3,1} \mathbf{k}_1 + a_{3,2} \mathbf{k}_2)), \quad a_{3,1} + a_{3,2} = c_3, \\ &\vdots \\ \mathbf{k}_\nu &= \mathbf{f}\left(t_n + c_\nu h, \mathbf{y}_n + h \sum_{j=1}^{\nu-1} a_{\nu,j} \mathbf{k}_j\right), \quad \sum_{j=1}^{\nu-1} a_{\nu,j} = c_\nu, \\ \mathbf{y}_{n+1} &= \mathbf{y}_n + h \sum_{l=1}^{\nu} b_l \mathbf{k}_l. \end{aligned}$$

The choice of the *RK coefficients* $a_{l,j}$ is motivated at the first instance by order considerations.

Example Set $\nu = 2$. We have $\mathbf{k}_1 = \mathbf{f}(t_n, \mathbf{y}_n)$ and, Taylor-expanding about (t_n, \mathbf{y}_n) ,

$$\begin{aligned} \mathbf{k}_2 &= \mathbf{f}(t_n + c_2 h, \mathbf{y}_n + h c_2 \mathbf{f}(t_n, \mathbf{y}_n)) \\ &= \mathbf{f}(t_n, \mathbf{y}_n) + h c_2 \left[\frac{\partial \mathbf{f}(t_n, \mathbf{y}_n)}{\partial t} + \frac{\partial \mathbf{f}(t_n, \mathbf{y}_n)}{\partial \mathbf{y}} \mathbf{f}(t_n, \mathbf{y}_n) \right] + \mathcal{O}(h^2). \end{aligned}$$

But

$$\mathbf{y}' = \mathbf{f}(t, \mathbf{y}) \quad \Rightarrow \quad \mathbf{y}'' = \frac{\partial \mathbf{f}(t, \mathbf{y})}{\partial t} + \frac{\partial \mathbf{f}(t, \mathbf{y})}{\partial \mathbf{y}} \mathbf{f}(t, \mathbf{y}).$$

Therefore, substituting the exact solution $\mathbf{y}_n = \mathbf{y}(t_n)$, we obtain $\mathbf{k}_1 = \mathbf{y}'(t_n)$ and $\mathbf{k}_2 = \mathbf{y}'(t_n) + h c_2 \mathbf{y}''(t_n) + \mathcal{O}(h^2)$. Consequently, the *local error* is

$$\begin{aligned} \mathbf{y}(t_{n+1}) - (\mathbf{y}(t_n) + h b_1 \mathbf{k}_1 + h b_2 \mathbf{k}_2) &= [\mathbf{y}(t_n) + h \mathbf{y}'(t_n) + \tfrac{1}{2} h^2 \mathbf{y}''(t_n) + \mathcal{O}(h^3)] \\ &\quad - [\mathbf{y}(t_n) + h(b_1 + b_2) \mathbf{y}'(t_n) + h^2 b_2 c_2 \mathbf{y}''(t_n) + \mathcal{O}(h^3)]. \end{aligned}$$

We deduce that the RK method is of order 2 if $b_1 + b_2 = 1$ and $b_2 c_2 = \frac{1}{2}$. We can demonstrate that no such method may be of order ≥ 3 . To show this consider the ODE $y' = y$ with $y(0) = 1$ whose solution is $y(t) = e^t$. For this ODE we can write the local error explicitly: indeed we have $k_1 = f(t_n, y(t_n)) = e^{t_n}$ and $k_2 = f(t_n + c_2 h, y(t_n) + c_2 h k_1) = y(t_n) + c_2 h k_1 = e^{t_n} (1 + c_2 h)$. Then the local error is

$$\begin{aligned} y(t_{n+1}) - (y(t_n) + h b_1 k_1 + h b_2 k_2) &= e^{t_{n+1}} - e^{t_n} - e^{t_n} (h b_1 + h b_2 + h^2 b_2 c_2) \\ &= e^{t_n} (e^h - 1 - h(b_1 + b_2) - h^2(b_2 c_2)) \\ &= e^{t_n} \left(h(1 - b_1 - b_2) + h^2(1/2 - b_2 c_2) + \frac{h^3}{6} + \mathcal{O}(h^4) \right). \end{aligned}$$

¹Corrections and suggestions to these notes should be emailed to h.fawzi@damtp.cam.ac.uk.

We see that there is no choice of b_1, b_2, c_1, c_2 that will make the term h^3 vanish, and so the method cannot have order ≥ 3 .

General RK methods A general ν -stage *Runge-Kutta method* is

$$\mathbf{k}_l = \mathbf{f} \left(t_n + c_l h, \mathbf{y}_n + h \sum_{j=1}^{\nu} a_{l,j} \mathbf{k}_j \right) \quad \text{where} \quad \sum_{j=1}^{\nu} a_{l,j} = c_l, \quad l = 1, 2, \dots, \nu,$$

$$\mathbf{y}_{n+1} = \mathbf{y}_n + h \sum_{l=1}^{\nu} b_l \mathbf{k}_l.$$

Obviously, $a_{l,j} = 0$ for all $l \leq j$ yields the standard *explicit* RK. Otherwise, an RK method is said to be *implicit*.

4.4 Stiff equations

Consider the linear scalar system

$$\begin{cases} y' = \lambda y \\ y(0) = 1 \end{cases}$$

where $\lambda < 0$. The solution is $y(t) = e^{\lambda t}$ which decays to 0 as $t \rightarrow \infty$. If we solve our ODE using a numerical method, we would like our sequence (y_n) to also decay to zero. For example with Euler's method we get $y_{n+1} = y_n + h\lambda y_n = (1 + h\lambda)y_n$ whose solution is $y_n = (1 + h\lambda)^n$. Thus the sequence y_n converges to 0 as $n \rightarrow \infty$ provided that $|1 + h\lambda| < 1$, i.e., $h < 2/|\lambda|$. For large λ this can be a severe restriction on h : for example for $\lambda = -1000$ this implies $h < 2/1000 = 0.002$.

Consider now the implicit Euler method. Here we have $y_{n+1} = y_n + h\lambda y_{n+1}$ which gives $y_{n+1} = (1 - h\lambda)^{-1} y_n$ and so $y_n = (1 - h\lambda)^{-n}$ which converges to 0 for any choice of $h > 0$ (we assumed $\lambda < 0$)!

Definition Suppose that a numerical method, applied to $y' = \lambda y$, $y(0) = 1$, with constant h , produces the solution sequence $\{y_n\}_{n \in \mathbb{Z}^+}$. We call the set

$$\mathcal{D} = \{h\lambda \in \mathbb{C} : \lim_{n \rightarrow \infty} y_n = 0\}$$

the *linear stability domain* of the method. Noting that the set of $\lambda \in \mathbb{C}$ for which $y(t) \xrightarrow{t \rightarrow \infty} 0$ is the left half-plane $\mathbb{C}^- = \{z \in \mathbb{C} : \operatorname{Re} z < 0\}$, we say that the method is *A-stable* if $\mathbb{C}^- \subseteq \mathcal{D}$.

Example We have already seen that for the explicit Euler's method $y_n \rightarrow 0$ iff $|1 + h\lambda| < 1$, therefore $\mathcal{D} = \{z \in \mathbb{C} : |1 + z| < 1\}$ and the explicit Euler method is not A-stable. Moreover, solving $y' = \lambda y$ with the implicit Euler method we have seen that $y_n \rightarrow 0$ iff $|1 - h\lambda|^{-1} < 1$, therefore the linear stability domain is $\mathcal{D} = \{z \in \mathbb{C} : |1 - z| > 1\}$, hence the implicit Euler method is A-stable.