Mathematical Tripos Part IB: Lent 2020 Numerical Analysis – Lecture 6¹

We look at another example of application of the Peano kernel theorem.

Example We approximate a derivative by a linear combination of function values, $f'(0) \approx -\frac{3}{2}f(0) + 2f(1) - \frac{1}{2}f(2)$. Define $L(f) := f'(0) - [-\frac{3}{2}f(0) + 2f(1) - \frac{1}{2}f(2)]$ and it is easy to check that L(f) = 0 for $f \in \mathbb{P}_2[x]$. (Verify by trying $f(x) = 1, x, x^2$ and using linearity of L.) Thus, for $f \in \mathbb{C}^3[0, 2]$ we have

$$L(f) = \frac{1}{2} \int_{0}^{2} K(\theta) f'''(\theta) d\theta$$

with $K(\theta) = L(x \mapsto (x - \theta)_+^2)$. For fixed θ , let $g(x) := (x - \theta)_+^2$. Then

$$\begin{split} K(\theta) &= L(g) = g'(0) - \left[-\frac{3}{2}g(0) + 2g(1) - \frac{1}{2}g(2) \right] \\ &= 2(0-\theta)_+ - \left[-\frac{3}{2}(0-\theta)_+^2 + 2(1-\theta)_+^2 - \frac{1}{2}(2-\theta)_+^2 \right] \\ &= \begin{cases} 2\theta - \frac{3}{2}\theta^2, & 0 \leq \theta \leq 1, \\ \frac{1}{2}(2-\theta)^2, & 1 \leq \theta \leq 2, \\ 0, & \text{else.} \end{cases} \end{split}$$

One can verify that $\int_0^2 |K(\theta)| d\theta = \frac{2}{3}$. Consequently for any $f \in C^3[0,2]$ we have

$$|L(f)| \le \frac{1}{2!} \int_0^2 |K(\theta)f'''(\theta)| d\theta \le \frac{1}{2} ||f'''||_{\infty} \int_0^2 |K(\theta)| d\theta = \frac{1}{3} ||f'''||_{\infty},$$

where $||f'''||_{\infty} = \max_{x \in [0,2]} |f'''(x)|$.

4 Ordinary differential equations

We wish to approximate the exact solution of the ordinary differential equation (ODE)

$$\mathbf{y}' = \mathbf{f}(t, \mathbf{y}), \qquad t \ge 0, \tag{4.1}$$

where $\boldsymbol{y} \in \mathbb{R}^N$ and the function $\boldsymbol{f} : \mathbb{R} \times \mathbb{R}^N \to \mathbb{R}^N$ is sufficiently 'nice'. (In principle, it is enough for \boldsymbol{f} to be Lipschitz to ensure that the solution exists and is unique. Yet, for simplicity, we henceforth assume that \boldsymbol{f} is analytic: in other words, we are always able to expand locally into Taylor series.) The equation (4.1) is accompanied by the initial condition $\boldsymbol{y}(0) = \boldsymbol{y}_0$.

Our purpose is to approximate $\mathbf{y}_{n+1} \approx \mathbf{y}(t_{n+1})$, $n = 0, 1, \ldots$, where $t_m = mh$ and the time step h > 0 is small, from $\mathbf{y}_0, \mathbf{y}_1, \ldots, \mathbf{y}_n$ and equation (4.1).

4.1 One-step methods

A one-step method is a map $y_{n+1} = \varphi_h(t_n, y_n)$, i.e. an algorithm which allows y_{n+1} to depend only on t_n, y_n, h and the ODE (4.1).

The Euler method: We know \boldsymbol{y} and its slope \boldsymbol{y}' at t=0 and wish to approximate \boldsymbol{y} at t=h>0. The most obvious approach is to truncate $\boldsymbol{y}(h)=\boldsymbol{y}(0)+h\boldsymbol{y}'(0)+\frac{1}{2}h^2\boldsymbol{y}''(0)+\cdots$ at the h^2 term. Since $\boldsymbol{y}'(0)=\boldsymbol{f}(t_0,\boldsymbol{y}_0)$, this procedure approximates $\boldsymbol{y}(h)\approx\boldsymbol{y}_0+h\boldsymbol{f}(t_0,\boldsymbol{y}_0)$ and we thus set $\boldsymbol{y}_1=\boldsymbol{y}_0+h\boldsymbol{f}(t_0,\boldsymbol{y}_0)$. By the same token, we may advance from h to 2h by letting $\boldsymbol{y}_2=\boldsymbol{y}_1+h\boldsymbol{f}(t_1,\boldsymbol{y}_1)$. In general, we obtain the Euler method

$$y_{n+1} = y_n + h f(t_n, y_n), \qquad n = 0, 1, \dots$$
 (4.2)

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Convergence: Let $t^* > 0$ be given. We say that a method, which for every h > 0 produces the solution sequence $\boldsymbol{y}_n = \boldsymbol{y}_n(h), n = 0, 1, \dots, \lfloor t^*/h \rfloor$, converges if

$$\lim_{h \to 0} \max_{n=0,...,|t^*/h|} \| \boldsymbol{y}_n(h) - \boldsymbol{y}(nh) \| = 0,$$

where y(nh) is the evaluation at time t = nh of the exact solution of (4.1).

Theorem Suppose that f satisfies the Lipschitz condition: there exists $\lambda \geq 0$ such that

$$\|\boldsymbol{f}(t,\boldsymbol{v}) - \boldsymbol{f}(t,\boldsymbol{w})\| \le \lambda \|\boldsymbol{v} - \boldsymbol{w}\|, \qquad t \in [0,t^*], \quad \boldsymbol{v},\boldsymbol{w} \in \mathbb{R}^N.$$

Then the Euler method (4.2) converges.

Proof. Let $e_n = y_n - y(t_n)$, the error at step n, where $0 \le n \le t^*/h$. Thus,

$$e_{n+1} = y_{n+1} - y(t_{n+1}) = [y_n + hf(t_n, y_n)] - [y(t_n) + hy'(t_n) + O(h^2)].$$

By the Taylor theorem, the $\mathcal{O}(h^2)$ term can be bounded uniformly for all $[0, t^*]$ in the underlying norm $\|\cdot\|$ by ch^2 , where c > 0 (Indeed if we take $c = \frac{1}{2} \max_{t \in [0, t^*]} \|\boldsymbol{y}''(t)\|$, then by Taylor's formula with integral remainder we get that for any t, h such that $0 \le t < t + h \le t^*$, $\|\boldsymbol{y}(t+h) - (\boldsymbol{y}(t) + h\boldsymbol{y}'(t))\| \le ch^2$.) Thus, using (4.1) and the triangle inequality,

$$\|e_{n+1}\| \le \|y_n - y(t_n)\| + h\|f(t_n, y_n) - f(t_n, y(t_n))\| + ch^2$$

 $\le \|y_n - y(t_n)\| + h\lambda\|y_n - y(t_n)\| + ch^2 = (1 + h\lambda)\|e_n\| + ch^2.$

Consequently, by induction,

$$\|e_{n+1}\| \le (1+h\lambda)^m \|e_{n+1-m}\| + ch^2 \sum_{j=0}^{m-1} (1+h\lambda)^j, \qquad m = 0, 1, \dots, n+1.$$

In particular, letting m = n + 1 and bearing in mind that $e_0 = 0$, we have

$$\|e_{n+1}\| \le ch^2 \sum_{i=0}^n (1+h\lambda)^j = ch^2 \frac{(1+h\lambda)^{n+1}-1}{(1+h\lambda)-1} \le \frac{ch}{\lambda} (1+h\lambda)^{n+1}.$$

For small h > 0 it is true that $0 < 1 + h\lambda \le e^{h\lambda}$. This and $(n+1)h \le t^*$ imply that $(1+h\lambda)^{n+1} \le e^{t^*\lambda}$, therefore $\|e_n\| \le \frac{ce^{t^*\lambda}}{\lambda}h \xrightarrow{h\to 0} 0$ uniformly for $0 \le nh \le t^*$ and the theorem is true.