

## Chapter 2 (Initial Value Problems) and Chapter 3 (Numerical Linear Algebra)

Outline solutions to homework assignment: Exercises 2.7, 2.14, 3.12 and 3.15.

**Exercise 2.7** (\*). In his seminal paper of 1901, Carl Runge gave an example of what we now call a *Runge-Kutta 2 method*, where

$$\Phi(t_i, y_i; h) = \frac{1}{4}f(t_i, y_i) + \frac{3}{4}f\left(t_i + \frac{2}{3}h, y_i + \frac{2}{3}hf(t_i, y_i)\right).$$

- (i) Show that it is consistent.
- (ii) Show how this method fits into the general framework of RK2 methods. That is,
  - (a) What are  $\alpha$ ,  $b$ ,  $\alpha$ , and  $\beta$ ?
  - (b) Do they satisfy the conditions

$$\beta = \alpha, \quad b = \frac{1}{2\alpha}, \quad a = 1 - b?$$

- (iii) Use it to estimate the solution at the point  $t = 2$  to  $y(1) = 1$ ,  $y' = 1 + t + y/t$  taking  $n = 2$  time steps.

**SOLUTION:**

- (i) The method is consistent if  $\Phi(t_i, y_i; 0) = f(t_i, y_i)$  (that is, if we formally set  $h = 0$  in the method, we recover the right-hand side of the differential equation). In this case,

$$\Phi(t_i, y_i; 0) = \frac{1}{4}f(t_i, y_i) + \frac{3}{4}f\left(t_i + 0 \cdot \frac{2}{3}, y_i + 0 \cdot \frac{2}{3}f(t_i, y_i)\right) = \frac{1}{4}f(t_i, y_i) + \frac{3}{4}f(t_i, y_i) = f(t_i, y_i),$$

as required.

- (ii) All RK-2 methods fall into the following framework:

$$k_1 = f(t_i, y_i), \quad k_2 = f(t_i + \alpha h, y_i + \beta h k_1), \quad \Phi(t_i, y_i; h) = \alpha k_1 + b k_2.$$

So we can see that  $\alpha = 1/4$  and  $b = 3/4$ . Furthermore, we see that  $k_2 = f(t_i + \frac{2}{3}h, y_i + \frac{2}{3}h k_1)$ . That is,  $\alpha = \beta = 2/3$ . So, to answer the question,

- (a)  $\alpha = 1/4$  and  $b = 3/4$ ,  $\alpha = \beta = 2/3$ .
- (b) Yes, since  $\alpha + b = (1 + 3)/4 = 1$  and  $\alpha = \beta$  and  $1/(2\alpha) = 3/4 = b$  as required.

- (iii) For this problem,  $f(x, t) = 1 + t + y/t$ ,  $t_0 = 1$  and  $y_0 = 1$ . Proceed as follows.

Step 1: To solve for  $y_1 \approx y(3/2)$ , we compute

$$k_1 = f(t_0, y_0) = f(1, 1) = 3, \quad \text{and} \quad k_2 = f\left(1 + \frac{1}{3}, 1 + \frac{1}{3}f(1, 1)\right) = f\left(\frac{4}{3}, 2\right) = \frac{23}{6}.$$

Then

$$y_1 = y_0 + h(\alpha k_1 + b k_2) = 1 + \frac{1}{2}\left(\frac{1}{4} \cdot 3 + \frac{3}{4} \cdot \frac{23}{6}\right) = \frac{45}{16}$$

Step 2: To Solve for  $y_2 \approx y(2)$ , we compute

$$k_1 = f\left(\frac{3}{2}, \frac{45}{16}\right) = \frac{35}{8}, \quad \text{and} \quad k_2 = f\left(\frac{3}{2} + \frac{1}{3}, \frac{45}{16} + \frac{1}{3} \cdot \frac{35}{8}\right) = f\left(\frac{11}{6}, \frac{205}{48}\right) = \frac{1363}{264}.$$

Then

$$y_2 = \frac{45}{16} + \frac{1}{2}\left(\frac{1}{4} \cdot \frac{35}{8} + \frac{3}{4} \cdot \frac{1363}{264}\right) = \frac{233}{44} \approx 5.2955.$$

Here are some entries for 3-stage Runge-Kutta method tableaux for Exercise 2.14.

**Method 0:**  $\alpha_2 = 2/3$ ,  $\alpha_3 = 0$ ,  $b_1 = 1/12$ ,  $b_2 = 3/4$ ,  $\beta_{32} = 3/2$

**Method 1:**  $\alpha_2 = 1/4$ ,  $\alpha_3 = 1$ ,  $b_1 = -1/6$ ,  $b_2 = 8/9$ ,  $\beta_{32} = 12/5$

**Method 2:**  $\alpha_2 = 1/4$ ,  $\alpha_3 = 1/2$ ,  $b_1 = 2/3$ ,  $b_2 = -4/3$ ,  $\beta_{32} = 2/5$

**Method 3:**  $\alpha_2 = 1/4$ ,  $\alpha_3 = 1/3$ ,  $b_1 = 3/2$ ,  $b_2 = -8$ ,  $\beta_{32} = 4/45$

**Method 4:**  $\alpha_2 = 1$ ,  $\alpha_3 = 1/4$ ,  $b_1 = -1/6$ ,  $b_2 = 5/18$ ,  $\beta_{32} = 3/16$

**Method 5:**  $\alpha_2 = 1$ ,  $\alpha_3 = 1/5$ ,  $b_1 = -1/3$ ,  $b_2 = 7/24$ ,  $\beta_{32} = 4/25$

**Method 6:**  $\alpha_2 = 1$ ,  $\alpha_3 = 1/6$ ,  $b_1 = -1/2$ ,  $b_2 = 3/10$ ,  $\beta_{32} = 5/36$

**Method 7:**  $\alpha_2 = 1/2$ ,  $\alpha_3 = 1/7$ ,  $b_1 = 7/6$ ,  $b_2 = 22/15$ ,  $\beta_{32} = -10/49$

**Method 8:**  $\alpha_2 = 1/2$ ,  $\alpha_3 = 1/8$ ,  $b_1 = 4/3$ ,  $b_2 = 13/9$ ,  $\beta_{32} = -3/16$

**Method 9:**  $\alpha_2 = 1/3$ ,  $\alpha_3 = 1/9$ ,  $b_1 = 4$ ,  $b_2 = 15/4$ ,  $\beta_{32} = -2/27$

**Exercise 2.14** (Your own RK3 method  $\star$ ). Answer the following questions for Method K from the list above, where K is the last digit of your ID number. For example, if your ID number is 01234567, use Method 7.

(a) Assuming that the method is consistent, determine the value of  $b_3$ .

(b) Consider the initial value problem:

$$y(0) = 1, \quad y'(t) = \lambda y(t).$$

Using that the solution is  $y(t) = e^{\lambda t}$ , write out a Taylor series for  $y(t_{i+1})$  about  $y(t_i)$  up to terms of order  $h^4$  (use that  $h = t_{i+1} - t_i$ ).

Using that your method should agree with the Taylor Series expansion up to terms of order  $h^3$ , determine  $\beta_{21}$  and  $\beta_{31}$ .

### SOLUTION:

(a) A method is consistent if

$$\Phi(t_i, y_i; 0) = f(t_i, y_i),$$

For all these methods, RK3 if  $h = 0$ , then  $k_1 = k_2 = k_3$ . Since

$$\Phi(t_i, y_i; h) = b_1 k_1 + b_2 k_2 + b_3 k_3,$$

it follows that the method is consistent if  $b_1 + b_2 + b_3 = 1$ . You can use this to find your  $b_3$

(b) The Taylor series expansion up to order  $h^4$  is

$$y(t_{i+1}) = y(t_i) + (t_{i+1} - t_i)y'(t_i) + \frac{(t_{i+1} - t_i)^2}{2!}y''(t_i) + \frac{(t_{i+1} - t_i)^3}{3!}y'''(t_i) + \frac{(t_{i+1} - t_i)^4}{4!}y''''(t_i).$$

Letting  $y_i$  approximate  $y(t_i)$ , and using that  $y^{(k)}(t) = \lambda^k y(t)$ , and  $t_{i+1} - t_i = h$ , we get

$$y(t_{i+1}) = y(t_i) \left( 1 + h\lambda + \frac{h^2\lambda^2}{2} + \frac{h^3\lambda^3}{6} \right) + \mathcal{O}(h^4).$$

Next, we write out the Runge-Kutta 3 method:

$$k_1 = f(t_i + \alpha_1 h, y_i) = \lambda y_i,$$

$$k_2 = f(t_i + \alpha_2 h, y_i + \beta_{21} h k_1) = \lambda(y_i + \beta_{21} h \lambda y_i),$$

$$k_3 = f(t_i + \alpha_3 h, y_i + \beta_{31} h k_1 + \beta_{32} h k_2) = \lambda((y_i + \beta_{31} h \lambda y_i) + \beta_{32} h \lambda(y_i + \beta_{21} h \lambda y_i)).$$

The method is

$$y_{i+1} = y_i + h(b_1 k_1 + b_2 k_2 + b_3 k_3)$$

So

$$\begin{aligned} y_{i+1} &= y_i + h \left( b_1 \lambda y_i + b_2 \lambda (y_i + \beta_{21} h \lambda y_i) + b_3 \lambda ((y_i + \beta_{31} h \lambda y_i) + \beta_{32} h \lambda (y_i + \beta_{21} h \lambda y_i)) \right) \\ &= y_i (1 + h \lambda (b_1 + b_2 + b_3) + h^2 \lambda^2 (b_2 \beta_{21} + b_3 (\beta_{31} + \beta_{32})) + h^3 \lambda^3 (b_3 \beta_{32} \beta_{21})) \end{aligned}$$

For this to agree with the Taylor series expansion, we would need

(i)  $(b_1 + b_2 + b_3) = 1$ , which we have already.

(ii)  $b_2 \beta_{21} + b_3 (\beta_{31} + \beta_{32}) = \frac{1}{2}$ , and

(iii)  $b_3 \beta_{32} \beta_{21} = \frac{1}{6}$ .

Since we already have  $b_3$  and  $\beta_{32}$  we can use (iii) to calculate  $\beta_{21}$ , and then substitute this into (ii) to find  $\beta_{31}$ .

Table of coefficients			
Method	$b_3$	$\beta_{21}$	$\beta_{31}$
0	1/6	2/3	-3/2
1	5/18	1/4	-7/5
2	5/3	1/4	1/10
3	15/2	1/4	11/45
4	8/9	1	1/16
5	25/24	1	1/25
6	6/5	1	1/36
7	-49/30	1/2	17/49
8	-16/9	1/2	5/16
9	-27/4	1/3	5/27

**Exercise 3.12** (\*). Show that, for any vector  $\mathbf{x} \in \mathbb{R}^n$ ,  $\|\mathbf{x}\|_\infty \leq \|\mathbf{x}\|_2$  and  $\|\mathbf{x}\|_2^2 \leq \|\mathbf{x}\|_1 \|\mathbf{x}\|_\infty$ . For each of these inequalities, give an example for which the equality holds. Deduce that  $\|\mathbf{x}\|_\infty \leq \|\mathbf{x}\|_2 \leq \|\mathbf{x}\|_1$ .

**Solution:** First recall the definitions of these norms:

$$\|\mathbf{x}\|_\infty = \max_{i=1, \dots, n} |x_i|, \quad \|\mathbf{x}\|_1 = \sum_{i=1}^n |x_i|, \quad \text{and} \quad \|\mathbf{x}\|_2 = \left( \sum_{i=1}^n x_i^2 \right)^{\frac{1}{2}}.$$

To show that  $\|\mathbf{x}\|_\infty \leq \|\mathbf{x}\|_2$ , let  $\mathbf{x} \in \mathbb{R}^n$  be the vector  $\mathbf{x} = (x_1, x_2, \dots, x_n)^T$ , where  $x_j$  is such that  $|x_j| \geq |x_i|$  for all  $i$ . Then,

$$\|\mathbf{x}\|_\infty^2 = x_j^2 \leq x_j^2 + (x_1^2 + x_2^2 + \dots + x_{j-1}^2 + \dots + x_{j+1}^2 + \dots + x_n^2) = \|\mathbf{x}\|_2^2,$$

where here we have used that each  $x_i^2 \geq 0$ .

Next, to show that  $\|\mathbf{x}\|_2^2 \leq \|\mathbf{x}\|_1 \|\mathbf{x}\|_\infty$  we will use that

$$\|\mathbf{x}\|_2^2 = x_1 x_1 + x_2 x_2 + \dots + x_j x_j + \dots + x_n x_n \leq (|x_1| + |x_2| + \dots + |x_j| + \dots + |x_n|) |x_j| = \|\mathbf{x}\|_1 \|\mathbf{x}\|_\infty$$

where again we use that  $\|\mathbf{x}\|_\infty = |x_j|$ .

There are examples where equality holds. E.g., if  $\mathbf{x} = (1, 0, \dots, 0)^T$ , then  $\|\mathbf{x}\|_\infty = \|\mathbf{x}\|_2 = \|\mathbf{x}\|_1 = 1$ . So, for this  $\mathbf{x}$ ,  $\|\mathbf{x}\|_\infty = \|\mathbf{x}\|_2$  and  $\|\mathbf{x}\|_2^2 = \|\mathbf{x}\|_1 \|\mathbf{x}\|_\infty$ .

To finish, we need to show that  $\|\mathbf{x}\|_2 \leq \|\mathbf{x}\|_1$ . Clearly this is true for  $\|\mathbf{x}\| = 0$ . Otherwise, Combine  $\|\mathbf{x}\|_\infty \leq \|\mathbf{x}\|_2$ , and  $\|\mathbf{x}\|_2^2 \leq \|\mathbf{x}\|_1 \|\mathbf{x}\|_\infty$ , to get that

$$\|\mathbf{x}\|_2^2 \leq \|\mathbf{x}\|_1 \|\mathbf{x}\|_2,$$

and so  $\|\mathbf{x}\|_2 \leq \|\mathbf{x}\|_1$ .

**Exercise 3.15** (\*). Prove that  $\|A\|_1 = \max_{j=1,\dots,n} \sum_{i=1}^n |a_{i,j}|$ . Hint: Suppose that  $\sum_{i=1}^n |a_{i,j}| \leq C$ , for  $j = 1, 2, \dots, n$ , and show that for any vector  $\mathbf{x} \in \mathbb{R}^n$   $\sum_{i=1}^n |(A\mathbf{x})_i| \leq C\|\mathbf{x}\|_1$ . Now find a vector  $\mathbf{x}$  such that  $\sum_{i=1}^n |(A\mathbf{x})_i| = C\|\mathbf{x}\|_1$ . Now deduce the result.

**SOLUTION:** For a given  $A \in \mathbb{R}^{n \times n}$  let

$$C_j = \sum_{i=1}^n |a_{i,j}|, \quad \text{and let } C = \max_{j=1,\dots,n} C_j.$$

That is,  $C$  is the sum of the largest column. Let  $\mathbf{x}$  be the vector  $\mathbf{x} = (x_1, x_2, \dots, x_n)^T$ . Then

$$\|A\mathbf{x}\|_1 = \sum_{i=1}^n |(A\mathbf{x})_i| = \sum_{i=1}^n \left| \sum_{j=1}^n a_{i,j}x_j \right| \leq \sum_{i=1}^n \sum_{j=1}^n |a_{i,j}| |x_j| = \sum_{j=1}^n |x_j| \sum_{i=1}^n |a_{i,j}| \leq \sum_{j=1}^n |x_j| C = C\|\mathbf{x}\|_1.$$

That is, for any  $\mathbf{x}$  whatsoever,  $\|A\mathbf{x}\|_1 / \|\mathbf{x}\|_1 \leq C$ . It follows, since  $\mathbf{x}$  is arbitrary that

$$\|A\|_1 := \max_{\mathbf{x} \in \mathbb{R}/\{0\}} \frac{\|A\mathbf{x}\|_1}{\|\mathbf{x}\|_1} \leq C.$$

To get equality, we need to choose a  $\mathbf{x}$  for which  $\|A\mathbf{x}\|_1 / \|\mathbf{x}\|_1 = C$ . Let  $q$  be such that column  $q$  of  $A$  has the largest sum. That is,  $C_q = C$ . Now let  $\mathbf{x} = \mathbf{e}^{(q)}$  (i.e., the vector whose only non-zero entry is  $x_q = 1$ ; equivalently, it is column  $q$  of the identity matrix). For this vector  $\|\mathbf{x}\|_1 = 1$ . Moreover,  $A\mathbf{x} = (a_{1,q}, a_{2,q}, \dots, a_{n,q})^T$ ; that is, it is the vector which is column  $q$  of  $A$ . So,  $\|A\mathbf{x}\|_1 = C$ . This gives, for this  $\mathbf{x}$ ,

$$\frac{\|A\mathbf{x}\|_1}{\|\mathbf{x}\|_1} = \frac{C\|\mathbf{x}\|_1}{\|\mathbf{x}\|_1} = C = \max_{j=1,\dots,n} \sum_{i=1}^n |a_{i,j}|.$$