Scientific Computing HW 3

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Problem 1. Pf. From the Butcher array,

$$A = \begin{bmatrix} \gamma & 0 \\ 1 - \gamma & \gamma \end{bmatrix}, \quad b = \begin{bmatrix} 1 - \gamma \\ \gamma \end{bmatrix}, \quad c = \begin{bmatrix} \gamma \\ 1 \end{bmatrix}$$

Check the 1st order accuracy condition.

$$\sum_{l=1}^{2} b_{l} = (1 - \gamma) + \gamma = 1$$

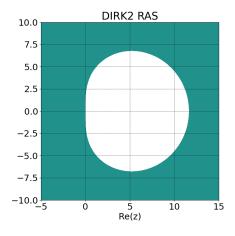
Check the 2nd order accuracy condition.

$$\sum_{l=1}^{2} b_l c_l = (1-\gamma)\gamma + \gamma \cdot 1 = \gamma - \gamma^2 + \gamma = 2\gamma - \gamma^2 = 2 - 2^{1/2} - 1 - 2^{-1} + 2^{1/2} = 1 - 2^{-1} = \frac{1}{2}$$

Thus the method is 2nd order accurate. To show it is A–stable, first find the stability function R(z) and let $|z| \to \infty$.

$$\begin{split} I - zA &= \begin{bmatrix} 1 - \gamma z & 0 \\ -(1 - \gamma)z & 1 - \gamma z \end{bmatrix} \implies D := \det(I - zA) = (1 - \gamma z)^2 = \gamma^2 z^2 - 2\gamma z + 1 \\ \implies (I - zA)^{-1} &= \frac{1}{D} \begin{bmatrix} 1 - \gamma z & 0 \\ (1 - \gamma)z & 1 - \gamma z \end{bmatrix} \implies (I - zA)^{-1} \mathbf{1}_{s \times 1} = \frac{1}{D} \begin{bmatrix} 1 - \gamma z \\ (1 - \gamma)z + 1 - \gamma z \end{bmatrix} = \frac{1}{D} \begin{bmatrix} 1 - \gamma z \\ (1 - 2\gamma)z + 1 \end{bmatrix} \\ R(z) - 1 &= zb^T (I - zA)^{-1} \mathbf{1}_{s \times 1} = \frac{z}{D} \left[(1 - \gamma)(1 - \gamma z) + \gamma((1 - 2\gamma)z + 1) \right] = \frac{z}{D} \left[1 - \gamma z - \gamma + \gamma^2 z + (\gamma - 2\gamma^2)z + \gamma \right] \\ \implies R(z) - 1 &= \frac{z}{D} \left[1 - \gamma^2 z \right] = \frac{-\gamma^2 z^2 + z}{\gamma^2 z^2 - 2\gamma z + 1} \implies R(z) = \frac{-\gamma^2 z^2 + z}{\gamma^2 z^2 - 2\gamma z + 1} + 1 \xrightarrow{|z| \to \infty} -1 + 1 = 0 \end{split}$$

To finish showing A-stability, we plot the RAS and see that it contains the left half plane. Code in 2nd cell of: https://github.com/RokettoJanpu/Scientific-Computing-2/blob/main/hw3%20RAS.ipynb



Problem 2. From the Butcher array,

$$A = \begin{bmatrix} \gamma & 0 \\ 1 - 2\gamma & \gamma \end{bmatrix}, b = \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix}, c = \begin{bmatrix} \gamma \\ 1 - \gamma \end{bmatrix}$$

1. **Pf.** Check the 1st order accuracy condition.

$$\sum_{l=1}^{2} b_l = \frac{1}{2} + \frac{1}{2} = 1$$

Check the 2nd order accuracy condition.

$$\sum_{l=1}^{2} b_l c_l = \frac{1}{2} \gamma + \frac{1}{2} (1 - \gamma) = \frac{1}{2} (\gamma + 1 - \gamma) = \frac{1}{2}$$

2. Pf. Check the 3rd order accuracy conditions.

$$\sum_{p,q,r} b_p a_{pq} a_{pr} = \frac{1}{2} [\gamma^2 + 2\gamma \cdot 0 + 0^2] + \frac{1}{2} [(1 - 2\gamma)^2 + 2(1 - 2\gamma)\gamma + \gamma^2]$$

The quantity in the second bracket is

$$(1 - 2\gamma)^2 + 2(1 - 2\gamma)\gamma + \gamma^2 = 1 + 4\gamma^2 - 4\gamma + 2\gamma - 4\gamma^2 + \gamma^2 = \gamma^2 - 2\gamma + 1 = (\gamma - 1)^2$$

giving

$$\sum_{p,q,r} b_p a_{pq} a_{pr} = \frac{1}{2} [\gamma^2 + \gamma^2 - 2\gamma + 1] = \gamma^2 - \gamma + \frac{1}{2}$$

We find

$$\gamma^2 = \frac{1}{2} + \frac{3}{36} + 2\frac{3^{1/2}}{12} = \frac{1}{12}[3 + 1 + 2 \cdot 3^{1/2}] = \frac{1}{12}[4 + 2 \cdot 3^{1/2}] = \frac{1}{6}[2 + 3^{1/2}]$$

so finally,

$$\sum_{p,q,r} b_p a_{pq} a_{pr} = \frac{1}{6} [2 + 3^{1/2} - 3 - 3^{1/2} + 3] = \frac{1}{3}$$

3. First find the stability function R(z).

$$\begin{split} I - zA &= \begin{bmatrix} 1 - \gamma z & 0 \\ -(1 - 2\gamma)z & 1 - \gamma z \end{bmatrix} \implies D := \det(I - zA) = (1 - \gamma z)^2 = \gamma^2 z^2 - 2\gamma z + 1 \\ &\implies (I - zA)^{-1} = \frac{1}{D} \begin{bmatrix} 1 - \gamma z & 0 \\ (1 - 2\gamma)z & 1 - \gamma z \end{bmatrix} \implies (I - zA)^{-1} \mathbf{1}_{s \times 1} = \frac{1}{D} \begin{bmatrix} 1 - \gamma z \\ (1 - 2\gamma)z + 1 - \gamma z \end{bmatrix} = \frac{1}{D} \begin{bmatrix} 1 - \gamma z \\ (1 - 3\gamma)z + 1 \end{bmatrix} \\ R(z) - 1 &= zb^T (I - zA)^{-1} \mathbf{1}_{s \times 1} = \frac{z}{2D} \left[1 - \gamma z + (1 - 3\gamma)z + 1 \right] = \frac{z}{2D} \left[(1 - 4\gamma)z + 2 \right] = \frac{1}{2} \frac{(1 - 4\gamma)z^2 + 2z}{\gamma^2 z^2 - 2\gamma z + 1} \end{split}$$

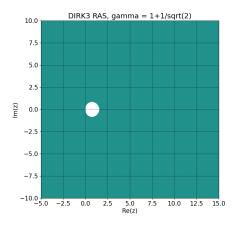
We find γ by imposing $\lim_{|z|\to\infty} R(z) = 0$.

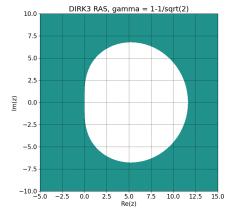
$$\lim_{|z| \to \infty} R(z) = 0 \iff -1 = \frac{1}{2} \lim_{|z| \to \infty} \frac{(1 - 4\gamma)z^2 + 2z}{\gamma^2 z^2 - 2\gamma z + 1} \iff \lim_{|z| \to \infty} \frac{(1 - 4\gamma)z^2 + 2z}{\gamma^2 z^2 - 2\gamma z + 1} = -2 \iff \frac{1 - 4\gamma}{\gamma^2} = -2$$

$$\iff -2\gamma^2 = 1 - 4\gamma \iff 2\gamma^2 - 4\gamma + 1 = 0 \iff \gamma = \frac{4}{4} \pm \frac{(16 - 8)^{1/2}}{4} = 1 \pm \frac{2 \cdot 2^{1/2}}{4} = 1 \pm 2^{-1/2}$$

We check that the method for $\gamma = 1 \pm 2^{-1/2}$ is A-stable, hence L-stable, by plotting the RASes and seeing that they contain the left half plane. Code in 3rd cell of:

https://github.com/RokettoJanpu/Scientific-Computing-2/blob/main/hw3%20RAS.ipynb





Problem 3.

1. Set $f(t,y) := -L(y - \phi(t)) + \phi'(t)$. To implement DIRK2, we first obtain explicit formulas for k_1, k_2, u_{n+1} .

$$k_{1} = f(t_{n} + \gamma h, u_{n} + h \gamma k_{1}) = -L \left[u_{n} + h \gamma k_{1} - \phi(t_{n} + \gamma h) \right] + \phi'(t_{n} + \gamma h)$$

$$\implies k_{1} = -Lh\gamma k_{1} - L \left[u_{n} - \phi(t_{n} + \gamma h) \right] + \phi'(t_{n} + \gamma h)$$

$$\implies (1 + Lh\gamma)k_{1} = -L \left[u_{n} - \phi(t_{n} + \gamma h) \right] + \phi'(t_{n} + \gamma h) \implies k_{1} = \frac{-L \left[u_{n} - \phi(t_{n} + \gamma h) \right] + \phi'(t_{n} + \gamma h)}{1 + Lh\gamma}$$

$$k_{2} = f(t_{n} + h, u_{n} + h(1 - \gamma)k_{1} + h\gamma k_{2}) = -L \left[u_{n} + h(1 - \gamma)k_{1} + h\gamma k_{2} - \phi(t_{n} + h) \right] + \phi'(t_{n} + h)$$

$$\implies k_{2} = -Lh\gamma k_{2} - L \left[u_{n} + h(1 - \gamma)k_{1} - \phi(t_{n} + h) \right] + \phi'(t_{n} + h)$$

$$\implies k_{2} = \frac{-L \left[u_{n} + h(1 - \gamma)k_{1} - \phi(t_{n} + h) \right] + \phi'(t_{n} + h)}{1 + Lh\gamma}$$

$$w_{n+1} = u_{n} + h \left[(1 - \gamma)k_{1} + \gamma k_{2} \right]$$

We do the same for DIRK3.

$$k_1 = f(t_n + \gamma h, u_n + h\gamma k_1)$$

so the explicit formula for k_1 is the same as in DIRK2.

$$k_{1} = \frac{-L\left[u_{n} - \phi(t_{n} + \gamma h)\right] + \phi'(t_{n} + \gamma h)}{1 + Lh\gamma}$$

$$k_{2} = f(t_{n} + (1 - \gamma)h, u_{n} + h(1 - 2\gamma)k_{2} + h\gamma k_{2})$$

$$= -L\left[u_{n} + h(1 - 2\gamma)k_{1} + h\gamma k_{2} - \phi(t_{n} + (1 - \gamma)h)\right] + \phi'(t_{n} + (1 - \gamma)h)$$

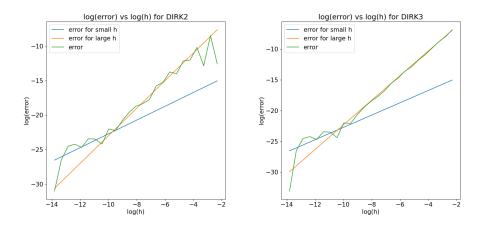
$$= -Lh\gamma k_{2} - L\left[u_{n} + h(1 - 2\gamma)k_{1} - \phi(t_{n} + (1 - \gamma)h)\right] + \phi'(t_{n} + (1 - \gamma)h)$$

$$\implies (1 + Lh\gamma)k_{2} = -L\left[u_{n} + h(1 - 2\gamma)k_{1} - \phi(t_{n} + (1 - \gamma)h)\right] + \phi'(t_{n} + (1 - \gamma)h)$$

$$\implies k_{2} = \frac{-L\left[u_{n} + h(1 - 2\gamma)k_{1} - \phi(t_{n} + (1 - \gamma)h)\right] + \phi'(t_{n} + (1 - \gamma)h)}{1 + Lh\gamma}$$

$$u_{n+1} = u_{n} + \frac{h}{2}[k_{1} + k_{2}]$$

 ${\bf Code:\ https://github.com/RokettoJanpu/Scientific-Computing-2/blob/main/hw3.ipynb}$



For larger values of h, about $\log h > -10$, the $\log -\log$ graph has a slope of 2, ie the error looks quadratic in h. For smaller values of h, about $\log h < -10$, the $\log -\log$ graph has a slope of 1, ie the error looks linear in h.