

2. EXERCISE SHEET

Due by: Monday, 29 April 2024, 12:00 noon (CEST) **(deadline extended)**

Please refer to **Assignment Submission Guideline** on Moodle

Modified on April 24, 2024

Problem 1. Implement the numerical model for the Lorenz-63 example (Example 1.1 in the textbook)

$$z^{n+1} = z^n + \delta t(f(z^n) + g(t_n)), \quad z^n = (x^n, y^n, z^n)^T,$$

with step-size $\delta t = 0.01$, initial condition $z^0 = (x^0, y^0, z^0) = (-0.587, -0.563, 16.870)^T$, $f(z) = (10(y - x), x(28 - z) - y, xy - \frac{8}{3}z)^T$, and $g(t_n) = g^n = (g_1^n, g_2^n, g_3^n)^T$ defined by

$$g_i^{n+1} = \begin{cases} 1.99999g_i^n + a/2 & \text{if } g_i^n \in [-a/2, 0), \\ -1.99999g_i^n + a/2 & \text{otherwise,} \end{cases} \quad i \in \{1, 2, 3\}, \quad n \geq 0,$$

where $(g_1^0, g_2^0, g_3^0) = (a(2^{-1/2} - 1/2), a(3^{-1/2} - 1/2), a(5^{-1/2} - 1/2))$ and $a = 1/\sqrt{\delta t}$.

(a) Plot the x-component of your solution over the time interval $[0, 10]$ similarly to Figure 1.3 in the textbook.

(b) Store the resulting reference trajectory in time intervals of $\Delta t_{\text{out}} = 0.05$ over 4000 cycles (i.e., between $t = 0$ and $t = 200$) in a file for later use in other examples. Print the mean and standard deviation of the three matrix rows corresponding to x , y , and z . (Do not store the system state from every single timestep as this becomes very inefficient, even for low dimensional problems; it is much better to overwrite the state vector on each timestep, and take a copy of the vector when you need to store it. The resulting data set should be stored in a matrix of size 3×4001 .)

Problem 2. Implement the numerical observation process as defined for the Lorenz-63 example using the reference trajectory generated in Exercise 1. That is, contaminate the observations with simulated noise via

$$x_{\text{obs}}(\mathbf{t}_k) = x(\mathbf{t}_k) + \frac{1}{20} \sum_{i=1+(k-1)20}^{20k} \xi_{10i}, \quad k = 1, \dots, 4000,$$

where $\mathbf{t}_k = k\Delta t_{\text{out}}$ for $k = 1, \dots, 4000$ and the process $(\xi_k)_{k \geq 0}$ is generated using the tent map iteration

$$\xi_{k+1} = \begin{cases} 1.99999\xi_k + a/2 & \text{if } \xi_k \in [-a/2, 0), \\ -1.99999\xi_k + a/2 & \text{otherwise,} \end{cases} \quad k \geq 0,$$

where $a = 4$ and $\xi_0 = a(2^{-1/2} - 1/2)$.

(a) Plot the observed x-components and the corresponding measurement errors over the time interval $[0, 10]$ similarly to Figure 1.3 in the textbook.

(b) Store the numerically generated observation values $y_{\text{obs}}(\mathbf{t}_k) = x_{\text{obs}}(\mathbf{t}_k)$ for $k = 1, \dots, N_{\text{obs}} = 4000$, in a file for later use. Print the mean and standard deviation of the vector $(y_{\text{obs}}(\mathbf{t}_k))_{k=1}^{4000}$.

Hint: You might obtain a trajectory different from the one displayed in the lecture notes. Differences can arise even for mathematically identical implementations due to round-off errors.

Problem 3. Follow the example on pages 11–12 of the lecture slides from week 2 (see also Example 1.4 in the textbook) and use linear extrapolation in order to produce forecasts for forecast intervals $\Delta t_{\text{out}} = 0.05$ and $3\Delta t_{\text{out}} = 0.15$, respectively, from the observations produced in Exercise 2. Plot your results over the time interval $[100, 105]$ as in the lecture slides. Compute the time averaged RMSE and discuss your findings.