

The Burden of Stability

Problem Statement

Consider the scalar ordinary differential equation (ODE):

$$\dot{x}(t) = f(t, x(t)) := \lambda x(t) + (1 - \lambda) \cos(t) - (1 + \lambda) \sin(t), \quad x(0) = 1 \quad (1)$$

The file `run_ODE.m` provides a framework to run various different methods for numerical integration of (1) with step-size $h > 0$ over time $t \in [0, T = 10]$. Let $N = \lfloor Th \rfloor$, where the notation $\lfloor \cdot \rfloor$ stands for the largest integer not exceeding z . Specifically, it computes a vector $(x_0 = x(0), x_1, \dots, x_N)$, where x_n 's are the proxies for $x(t_n)$ computed recursively via different methods with $tn = nh$. Define the average error of any numerical integration method applied to this ODE as

$$\mathcal{E} := \frac{1}{N+1} \sum_{n=0}^N |x(t_n) - x_n|$$

Use code or other methods to generate the plots required below and answer the questions. Please submit your code (at least for parts c and d).

a

Show that the analytical solution of the DOE is given by $x(t) = \cos(t) + \sin(t)$.

b

Plot the result of numerical integration via forward Euler method with $h = 0.15, 0.30, 0.45$ over $[0, T]$ together with the analytical solution. Comment how the average error for forward Euler \mathcal{E}_{FE} varies with h . Is forward Euler method stable for all values of h you simulated?

c

To implement backward Euler method for a given step-size $h > 0$, one needs to solve the nonlinear equation

$$x_{n+1} := x_n + hf(t_{n+1}, x_{n+1})$$

in each iteration $n \geq 0$. Implement Newton-Raphson to solve the equation $F(y) = 0$ where

$$F(y) := y - x_n - hf(t_{n+1}, y)$$

starting from the forward Euler solution, given by $y^{(0)} := x_n + hf(t_n, x_n)$. Iterate until $|F(y)| < 10^{-5}$. For the same values of h used in part (b), numerically integrate (1) using backward Euler method and plot the results together with the analytical solution on the same graph. Comment how the average error for backward Euler \mathcal{E}_{BE} varies with h . Is backward Euler method stable for all values of h you simulated?

d

To implement the trapezoidal method with step-size $h > 0$, one needs to solve the nonlinear equation

$$x_{n+1} := x_n + \frac{h}{2} [f(t_n, x_n) + f(t_{n+1}, x_{n+1})]$$

in each iteration $n \geq 0$. Implement Newton-Raphson to solve the equation $F(y) = 0$ with

$$F(y) := y - x_n - \frac{h}{2}[f(t_n, x_n) + f(t_{n+1}, y)]$$

starting from the forward Euler solution given by $y^{(0)} := x_n + hf(t_n, x_n)$. Iterate until $|F(y)| < 10^{-5}$. For the same values used in part (b), numerically integrate (1) using the trapezoidal method and plot the results together with the analytical solution on the same graph. Comment how the average error for this method varies with h . Is this method stable for the values of h you've simulated?

Solution

a

$$\begin{aligned}\dot{x} &= \lambda(\cos(t) + \sin(t)) + (1 - \lambda)\cos(t) - (1 + \lambda)\sin(t) \\ &= \lambda\cos(t) + (1 - \lambda)\cos(t) + \lambda\sin(t) - (1 + \lambda)\sin(t) \\ &= (\lambda + 1 - \lambda)\cos(t) + (\lambda - 1 - \lambda)\sin(t) \\ &= \cos(t) - \sin(t)\end{aligned}$$

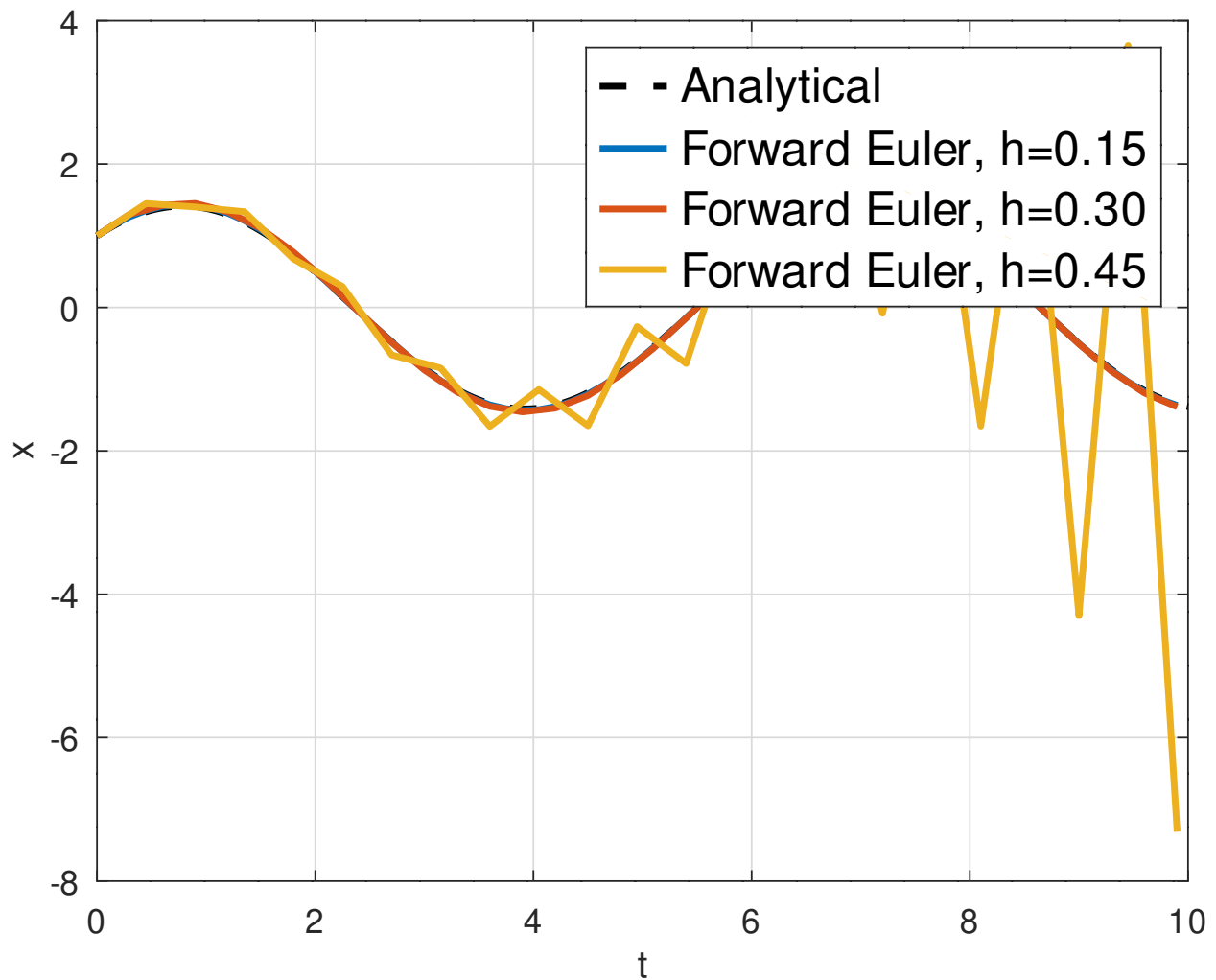
And indeed:

$$\frac{\partial}{\partial t}x(t) = \frac{\partial}{\partial t}(\cos(t) + \sin(t)) = -\sin(t) + \cos(t) = \cos(t) - \sin(t)$$

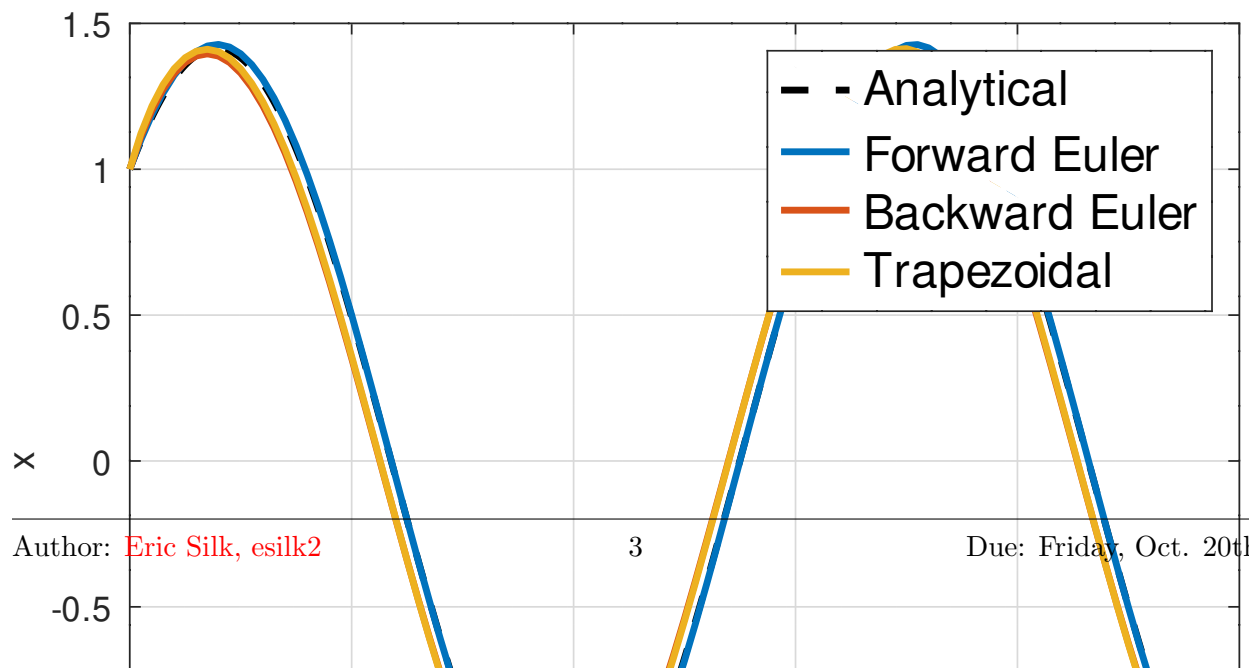
□

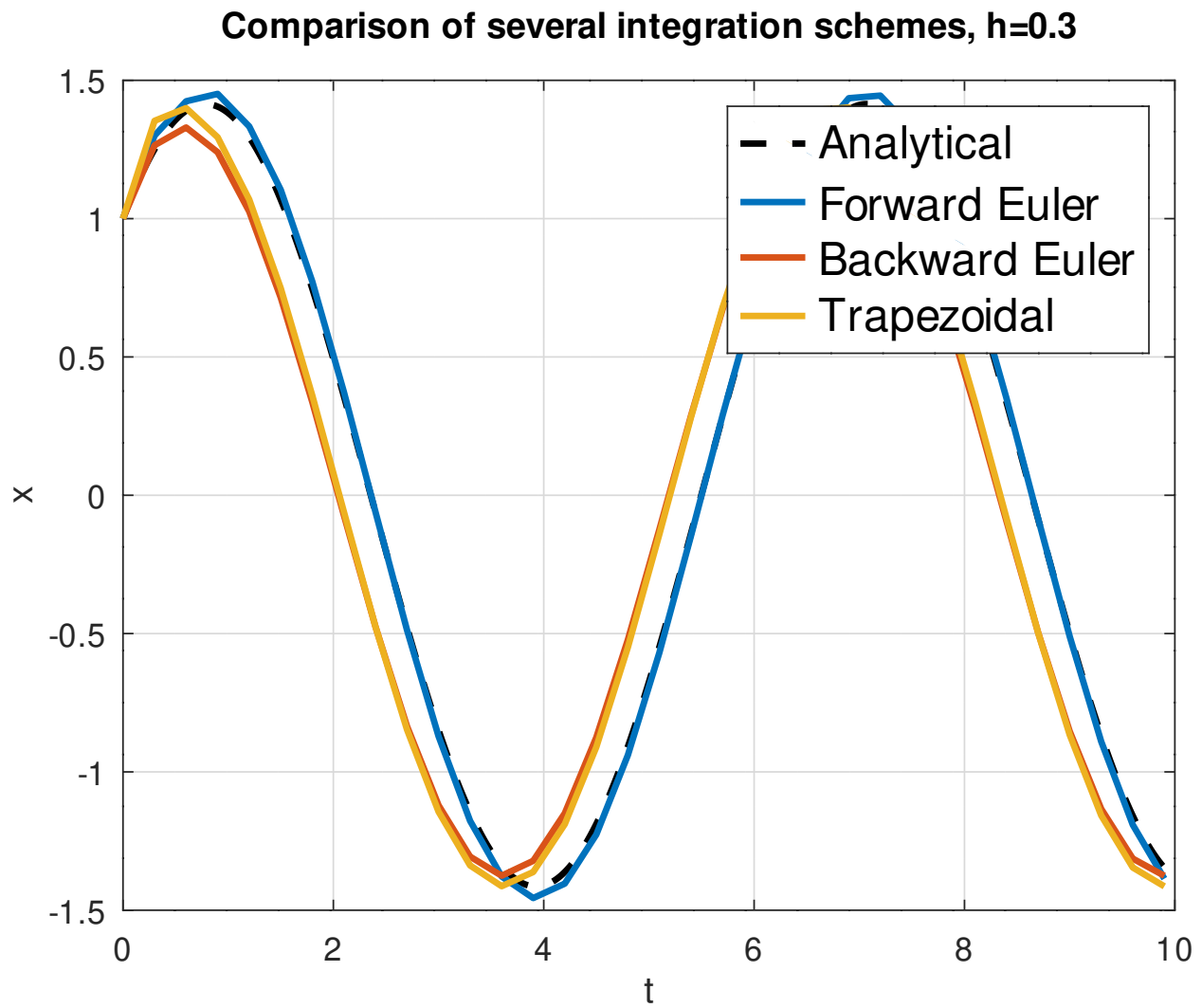
b

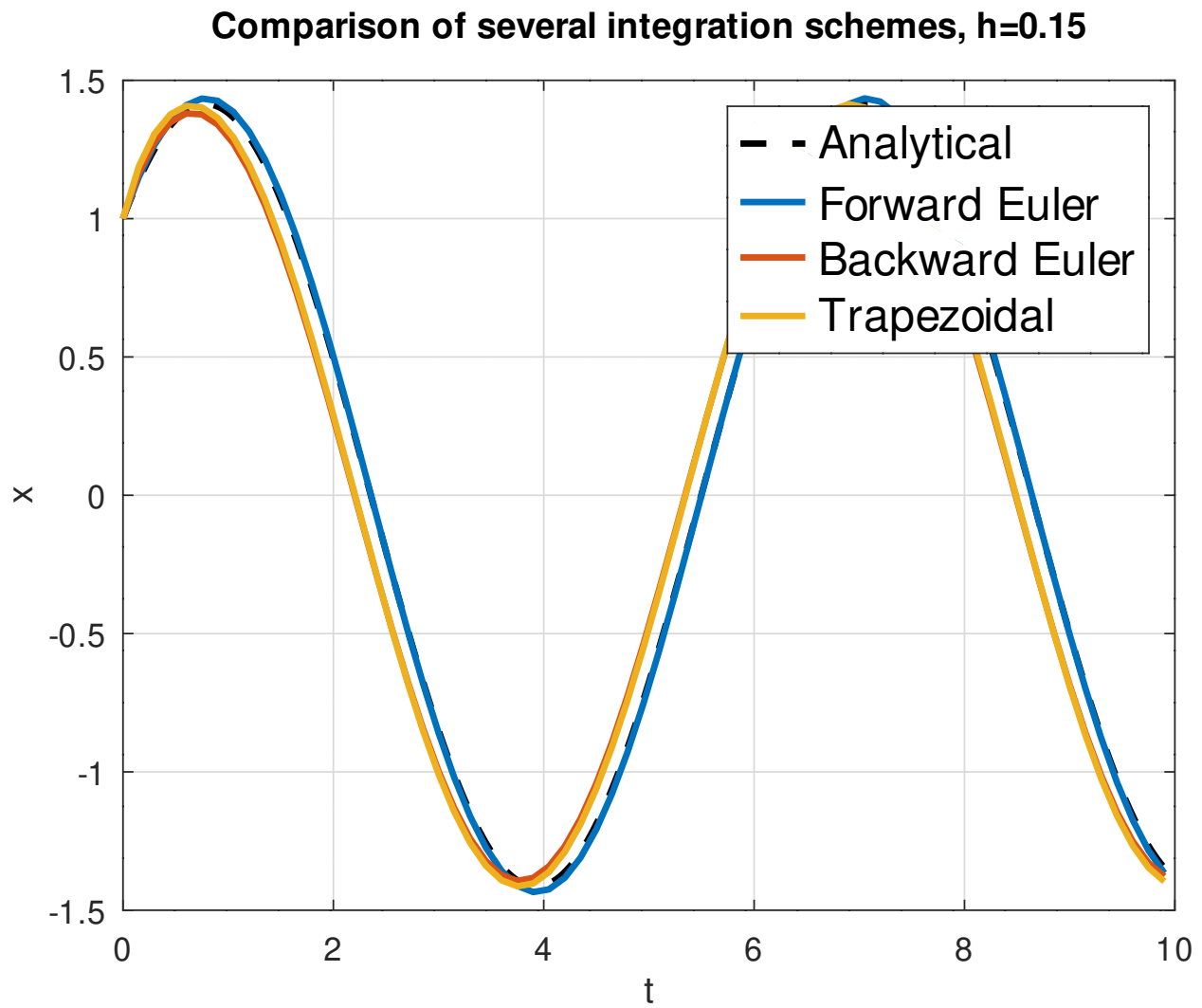
Mean error reported is 0.013344 for $h = 0.15$, 0.027145 for $h = 0.30$, and 1.2763 for $h = 0.45$. Also, it is visually apparent that the solution for $h = 0.45$ diverges, indicating numerical instability.

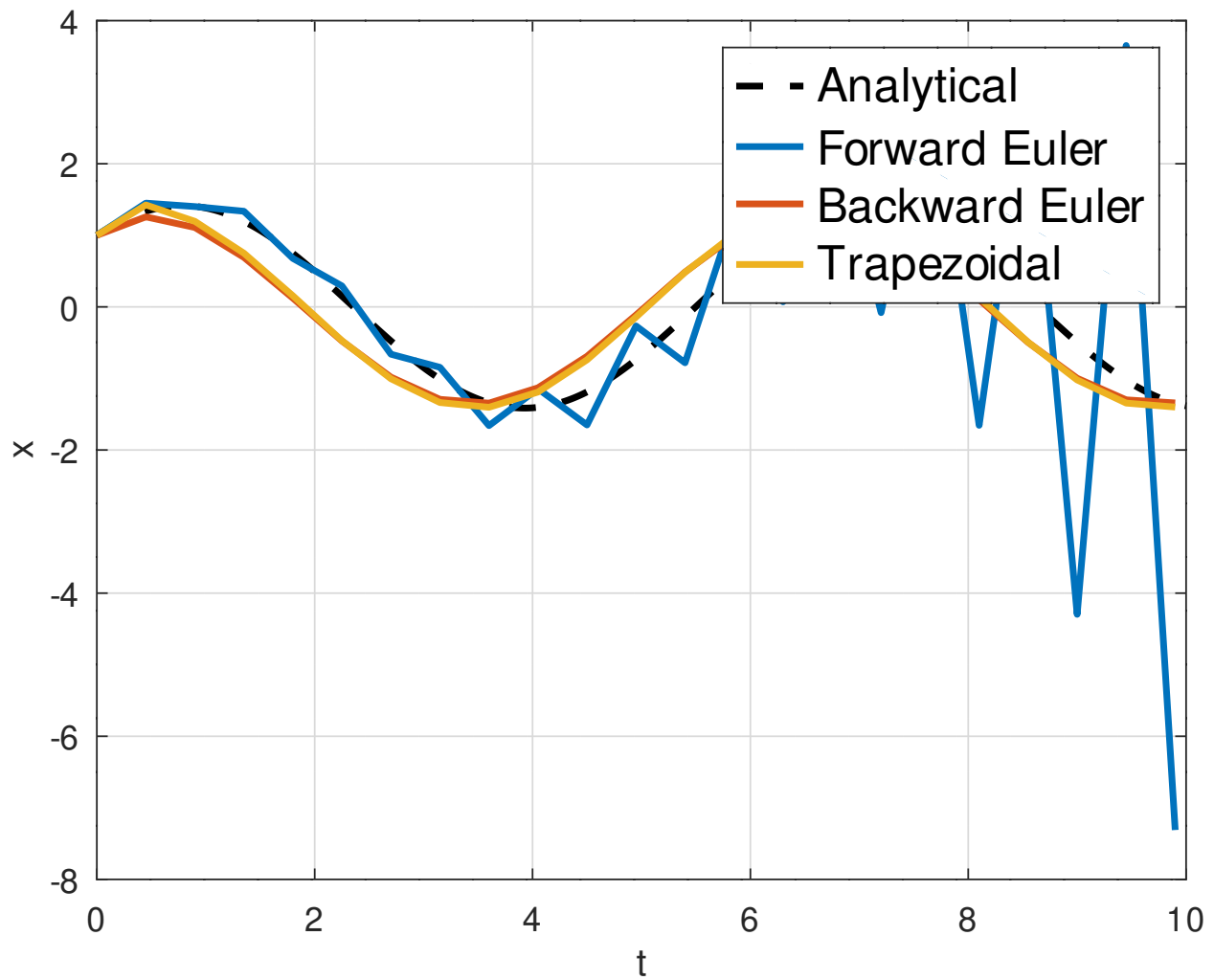
Comparison of several step sizes for Forward Euler integration

c, d

Comparison of several integration schemes, $h=0.1$ 





Comparison of several integration schemes, $h=0.45$ 

```
h=0.1
Processing analytical solution.
Processing forward Euler method.
Mean error =0.0088783
Processing backward Euler method.
Mean error =0.086862
Processing trapezoidal method.
Mean error =0.085593
```

```
h=0.15
Processing analytical solution.
Processing forward Euler method.
Mean error =0.013344
Processing backward Euler method.
Mean error =0.13035
Processing trapezoidal method.
Mean error =0.12856
```

```
h=0.3
Processing analytical solution.
Processing forward Euler method.
Mean error =0.027145
Processing backward Euler method.
Mean error =0.25538
Processing trapezoidal method.
Mean error =0.25064
```

```
h=0.45
Processing analytical solution.
Processing forward Euler method.
Mean error =1.2763
Processing backward Euler method.
Mean error =0.37076
Processing trapezoidal method.
Mean error =0.36708
```

Backward Euler and Trapezoidal both appear to be stable for all simulated values. Forward Euler is shown for completeness, and it blows up as expected.

Problem 2: Conjugacy is Independence

Problem Statement

If $\{d^0, \dots, d^{n-1}\}$ are pairwise Q -conjugate vectors in \mathbb{R}^n for $Q \succ 0$ then prove they are linearly independent.

Solution

Recall a sequence of vectors are linearly independent if:

$$\alpha_0 d_0 + \alpha_1 d_1 + \dots + \alpha_k d_k = 0$$

where $\{\alpha_1, \dots, \alpha_k\} \setminus 0 \neq \emptyset$ (i.e. there's at least one non-zero α).

If $\sum_{i=0}^{n-1} \alpha_i d_i$ then for $i_0 \in 0, 1, \dots, n-1$

$$0 = d_{i_0}^T Q \sum_{i=0}^{n-1} \alpha_i d_i = \alpha_{i_0} d_{i_0}^T Q d_{i_0}$$

So $\alpha_i = 0 \forall i = 0, \dots, n-1$. □

Credit to Dr. Burke's website[1].

Problem 3: The Price of Laziness

Problem Statement

For the ODE given by $\dot{x} = f(x, t)$, investigate the absolute stability of the following numerical integration schemes:

1. Trapezoidal rule with one step fixed-point iteration from forward Euler, given by:

$$x_{n+1} = x_n + \frac{h}{2}[f(x_n, t_n) + f(x_n + hf(x_n, t_n), t_{n+1})]$$

2. Backward Euler with one step fixed-point iteration from forward Euler, given by:

$$x_{n+1} = x_n + \frac{h}{2}f(x_n + hf(x_n, t_n), t_{n+1})$$

Solution

a

From the notes, let $f(x, t) := \lambda x$. Because this system is autonomous, we can simplify and say:

$$\begin{aligned} f(x_n, t_n) &= f(x_n) = \lambda x_n \\ f(x_n + hf(x_n, t_n), t_{n+1}) &= f(x_n + hf(x_n)) = f(x_n + h(\lambda x_n)) = \lambda(x_n + h(\lambda x_n)) \\ x_{n+1} &= x_n + \frac{h}{2}[\lambda x_n + \lambda(x_n + h\lambda x_n)] \\ x_{n+1} &= x_n(1 + \frac{h}{2}[\lambda + \lambda(1 + h\lambda)]) \\ x_{n+1} &= \frac{x_n}{2}(h^2\lambda^2 + 2h\lambda + 2) \\ \implies x_n &= (\frac{1}{2}h^2\lambda^2 + h\lambda + 1)^n \end{aligned}$$

Which, taking the limit as $n \rightarrow \infty$, we see that it will only go to zero for

$$\begin{aligned} -1 &< \frac{1}{2}h^2\lambda^2 + h\lambda + 1 < 1 \\ \implies -2 &< \frac{1}{2}(h\lambda)^2 + h\lambda < 0 \end{aligned}$$

Considering the left term:

$$0 < \frac{1}{2}(h\lambda)^2 + h\lambda + 2$$

which holds $\forall h\lambda$. The right portion, then:

$$\frac{1}{2}(h\lambda)^2 + h\lambda < 0$$

This is true for all $-2 < h\lambda < 0$. Recall $\lambda < 0$ and $h > 0$; therefore, $h\lambda < 0$ and the above equation is only restricted by the minimum value. Again, because λ is strictly negative, I'm going to relate this using the absolute value of λ (otherwise the negatives make the expression weird...):

$$h < 2/|\lambda|$$

Conceptually, the faster the prototypical exponential function decays (more negative λ), the smaller our step size needs to be. Because this is not stable $\forall h > 0$, we cannot say this is absolutely stable.

b

Proceeding as before:

$$x_{n+1} = x_n + h\lambda(x_n + h\lambda x_n) = x_n(1 + h\lambda + (h\lambda)^2)$$

So for stability:

$$|1 + h\lambda + (h\lambda)^2| < 1 \implies -1 < 1 + h\lambda + (h\lambda)^2 < 1$$

The left relationship is true $\forall h\lambda \in \mathbb{R} \implies \forall h \in \mathbb{R}$. The right, however, is only true for:

$$-1 < h\lambda < 0$$

Which does not hold $\forall h \in \mathbb{R}$. Therefore, the method is not absolutely stable.

Code

“run_ODE.m”

```

1  clc
2  clear
3  close all
4
5  % Code for HW 5, ECE 530, Fall 2023.
6
7  % Consider the ODE \dot{x} = f(t, x), starting from x0.
8  % Define the function f.
9  L = -5;
10 f = @(t, x) (L*x + (1-L) * cos(t) - (1 + L) * sin(t));
11
12 % Define the derivative of 'f' with respect to 'x'.
13 deriv_f = @(t,x) (L);
14 f_prime = deriv_f;
15
16 % Initial point
17 x0 = 1;
18
19 % Time horizon
20 T = 10;
21
22 hs = [0.1, 0.15, 0.30, 0.45];
23
24 for i=1:numel(hs)
25     % Step-size
26     h = hs(i);
27
28     % Number of iterations
29     N = floor(T/h);
30     times = (0:h:N*h)';
31
32     % Create a vector of results in x. Notice that our implementation is
33     % such that x(1) = x_0, x(2) = x_1, x(3) = x_2, etc.
34     x = zeros(1 + N, 1);
35     x(1) = x0;
36
37     figure()
38     %
39     % Analytical Solution of ODE
40     %
41     disp('=====')
42     disp(strcat("h=", num2str(h)))
43     disp('Processing analytical solution.')
```

```

87 % backward Euler method. Implement a Newton-Raphson method
88 % to compute x(n+1). Start the NR iteratiion from the explicit
89 % Euler solution. Iterate till | F(y) | > 10^{-5}
90
91 % Insert your code here to approximately solve F(y) = 0.
92 F = @(ynp1) ynp1-y-h*f(h*(n+1), ynp1);
93 F_prime = @(ynp1) 1-h*f_prime(h*(n+1), ynp1);
94 y = newton_raphson(y, F, F_prime, 1e-5, 1000);
95 x(n+1) = y;
96 end
97
98
99 % Plot the outcome.
100 plot(times, x, 'Linewidth', 2)
101 hold on
102
103 % Compute the average error.
104 display(strcat(...
105     'Mean error = ', ...
106     num2str(mean(abs(x - solution_ODE(times))))...
107 ))
108
109
110 %-----
111 % Trapezoidal method.
112 %-----
113
114 disp('Processing trapezoidal method.')
115 y = x(1);
116 for n = 1:N
117
118     % Define F(y) such that F(y)=0 is equivalent to solving
119     % the implicit equation that arises in each iteration of
120     % the trapezoidal method. Implement a Newton-Raphson method
121     % to compute x(n+1). Start the NR iteratiion from the explicit
122     % Euler solution. Iterate till | F(y) | > 10^{-5}
123
124     % Insert your code here to approximately solve F(y) = 0.
125     F = @(ynp1) ynp1-y-(h/2)*(f(h*n, y)+f(h*(n+1), ynp1));
126     F_prime = @(ynp1) 1-0.5*h*f_prime(h*(n+1), ynp1);
127     y = newton_raphson(y, F, F_prime, 1e-5, 1000);
128     x(n+1) = y;
129
130 end
131
132 % Plot the outcome.
133 plot(times, x, 'Linewidth', 2)
134 xlabel('t')
135 ylabel('x')
136 title(strcat("Comparison of several integration schemes, h=", num2str(h)))
137 hold on
138
139 % Compute the average error.
140 display(strcat(...
141     'Mean error = ', ...
142     num2str(mean(abs(x - solution_ODE(times))))...
143 ))
144
145 %-----
146 % Add legends, grid to the plot.
147 %-----
148
149 legend({'Analytical', 'Forward Euler', 'Backward Euler', 'Trapezoidal'}, 'FontSize', 14)
150 grid on
151 hold off
152 print(strcat("integration_", num2str(h), '.eps'), '-deps', '-color')
153 endfor

```

“forward_euler_stability.m”

```

1  clc
2  clear
3  close all
4
5  % Code for HW 5, ECE 530, Fall 2023.
6
7  % Consider the ODE  $\dot{x} = f(t, x)$ , starting from  $x_0$ .
8  % Define the function f.
9  L = -5;
10 f = @(t, x) (L*x + (1-L) * cos(t) - (1 + L) * sin(t));
11
12 % Define the derivative of 'f' with respect to 'x'.
13 deriv_f = @(t,x) (L);
14 f_prime = deriv_f;
15
16 % Initial point
17 x0 = 1;
18
19 % Time horizon
20 T = 10;
21
22 % Step-size
23 hs = [0.15, 0.3, 0.45];
24 h=hs(1);
25
26 % Number of iterations
27 N = floor(T/h);
28 times = (0:h:N*h)';
29
30 % Create a vector of results in x. Notice that our implementation is
31 % such that x(1) = x_0, x(2) = x_1, x(3) = x_2, etc.
32 x = zeros(1 + N, 1);
33 x(1) = x0;
34
35
36 %-----
37 % Analytical Solution of ODE
38 %-----
39
40 disp('Processing analytical solution.')
41
42 % Define the analytical solution.
43 solution_ODE = @(t) (cos(t) + sin(t));
44
45 % Plot the analytical solution.
46 times_cont = (0:0.01:T)';
47 figure(1);
48 plot(times_cont, solution_ODE(times_cont), 'k—', 'Linewidth', 2)
49 hold on
50
51
52 %-----
53 % Forward Euler method.
54 %-----
55
56 disp('Processing forward Euler method.')
57 for i = 1:3
58     h = hs(i);
59     N = floor(T/h);
60     times = (0:h:N*h)';
61
62     % Create a vector of results in x. Notice that our implementation is
63     % such that x(1) = x_0, x(2) = x_1, x(3) = x_2, etc.
64     x = zeros(1 + N, 1);
65     x(1) = x0;
66     % Implement the method.
67     for n = 1:N
68         x(n+1) = x(n) + h * f((n-1) * h, x(n));
69     end
70     % Plot the outcome.
71     plot(times, x, 'Linewidth', 2)
72     xlabel("t")
73     ylabel("x")
74     title("Comparison of several step sizes for Forward Euler integration")
75     hold on
76
77     % Compute the average error.
78     display(strcat(...
79         'Mean error = ', ...
80         num2str(mean(abs(x - solution_ODE(times)))...
81         ))
82 endfor
83
84
85 legend({'Analytical', 'Forward Euler, h=0.15', 'Forward Euler, h=0.30', 'Forward Euler, h=0.45'}, '
      FontSize',14)
86 grid on
87 hold off
88 print('forward_euler.eps', '-deps', '-color')

```

“newton_raphson.m”

```
1 function xn = newton_raphson(x0, func, deriv, tol, max_iter)
2   x = x0;
3   for n = 1:max_iter
4       fx = func(x);
5       if (abs(fx) < tol)
6           xn = x;
7           return;
8       endif
9       fprime = deriv(x);
10      dx = fx/fprime;
11      if isnan(dx)
12          display("NaN encountered, halting!")
13          xn = x;
14          return;
15      endif
16      tmp = x-dx;
17      if (isnan(tmp))
18          display("NaN would be produced, halting!")
19          xn = x;
20          return;
21      endif
22      x = x - dx;
23  endfor
24  xn = x0;
25  disp("Failed to converge!")
26 end
```

Bibliography

- [1] James V Burke. *Conjugate Direction Methods - University of Washington*. Feb. 2007. URL: <https://sites.math.washington.edu/~burke/crs/408f/notes/nlp/cg.pdf>.