

Homework Submission

CMDA 3605 Mathematical Modeling: Methods and Tools I

Virginia Tech Honor Code Pledge:

"I have neither given nor received unauthorized assistance on this assignment."

Solution to Exercise 1

(a)

$$\begin{aligned}\dot{S}(t) &= -\beta S(t)I(t) + \mu(P - S(t) - I(t)), \\ \dot{I}(t) &= \beta S(t)I(t) - \gamma I(t)\end{aligned}$$

Set equations equal to 0 to find steady states:

$$\begin{aligned}0 &= -\beta S(t)I(t) + \mu(P - S(t) - I(t)), \\ 0 &= \beta S(t)I(t) - \gamma I(t)\end{aligned}$$

Solve for second equation first:

$$\begin{aligned}\beta S(t)I(t) - \gamma I(t) &= 0 \\ \beta \hat{S} \hat{I} - \gamma \hat{I} &= 0 \\ \hat{I}(\beta \hat{S} - \gamma) &= 0\end{aligned}$$

Case 1: $\hat{I} = 0$. Solve for inside of parentheses, from the equation above we can tell that $\hat{I} = 0$. We will use the calculated value of \hat{S} later:

$$\begin{aligned}(\beta \hat{S} - \gamma) &= 0 \\ \beta \hat{S} &= \gamma \\ \hat{S} &= \frac{\gamma}{\beta}\end{aligned}$$

Now back to the first equation, substitute in $\hat{I} = 0$ back into the first equation to solve for \hat{S}_1 :

$$\begin{aligned}-\beta \hat{S} * 0 + \mu(P - \hat{S} - 0) &= 0 \\ \mu(P - \hat{S}) &= 0 \\ \hat{S} &= P\end{aligned}$$

This proves that $\hat{x}_1 = \begin{bmatrix} \hat{S}_1 \\ \hat{I}_1 \end{bmatrix} = \begin{bmatrix} P \\ 0 \end{bmatrix}$ is a steady state of the simplified SIR model.

Case 2: $\hat{S} = \frac{\gamma}{\beta}$ Substitute \hat{S} into the first equation to solve for \hat{I}_2 :

$$\begin{aligned}-\beta\left(\frac{\gamma}{\beta}\right)\hat{I} + \mu\left(P - \frac{\gamma}{\beta} - \hat{I}\right) &= 0 \\ -\gamma\hat{I} - \mu\left(P - \frac{\gamma}{\beta} - \hat{I}\right) &= 0 \\ -\gamma\hat{I} - \mu P - \mu\frac{\gamma}{\beta} - \mu\hat{I} &= 0 \\ (\gamma + \mu)\hat{I} &= \mu\left(P - \frac{\gamma}{\beta}\right) \\ \hat{I} &= \frac{\mu\left(P - \frac{\gamma}{\beta}\right)}{\gamma + \mu}\end{aligned}$$

Solution to Exercise 1 cont.

This proves that $\hat{x}_2 = \begin{bmatrix} \hat{S}_2 \\ \hat{I}_2 \end{bmatrix} = \begin{bmatrix} \frac{\gamma}{\beta} \\ \frac{\mu(P - \frac{\gamma}{\beta})}{\gamma + \mu} \end{bmatrix}$ is a steady state of the simplified SIR model.

Solution to Exercise 1 cont.

(b)

We need the Jaccobian for the stability analysis of the continuous time nonlinear system:

$$\begin{aligned} J(S, I) &= \begin{bmatrix} -\beta I - \mu & -\beta S - \mu \\ \beta I & \beta S - \gamma \end{bmatrix} \\ J(\hat{x}_1) &= \begin{bmatrix} -\beta(0) - \mu & -\beta P - \mu \\ \beta(0) & \beta(P) - \gamma \end{bmatrix} \\ &= \begin{bmatrix} -\mu & -\beta P - \mu \\ 0 & \beta(P) - \gamma \end{bmatrix} \end{aligned}$$

The eigenvalues can be read off the diagonal and are equal to:

- $\lambda_1 = -\mu$,
- $\lambda_2 = \beta P - \gamma$

Given that the re susceptibility rate μ will always be greater than 0, The following is always true:

$$\lambda_1 < 0$$

If $\lambda_2 < 0$, the steady state $\hat{x}_1 = \begin{bmatrix} P \\ 0 \end{bmatrix}$ will be asymptotically stable since:
spectral abscissa $\alpha(J[\hat{x}_1]) < 0$

additionally,

$$\lambda_2 = \beta P - \gamma \tag{1}$$

$$\tag{2}$$

and we know, $\lambda_2 < 0$

$$\beta P - \gamma < 0 \tag{3}$$

$$\beta P < \gamma \tag{4}$$

$$P < \frac{\gamma}{\beta} \tag{5}$$

So if $P < \frac{\gamma}{\beta}$, \hat{x}_1 becomes asymptotically stable.

Solution to Exercise 1 cont.

If $\lambda_2 > 0$ then the steady state $\hat{x}_1 = \begin{bmatrix} P \\ 0 \end{bmatrix}$ will be unstable since spectral abscissa $\alpha(J[\hat{x}_1]) > 0$

additionally,

$$\lambda_2 > 0 \tag{6}$$

$$\tag{7}$$

$$\beta P - \gamma > 0 \tag{8}$$

$$\beta P > \gamma \tag{9}$$

$$P > \frac{\gamma}{\beta} \tag{10}$$

So if $P > \frac{\gamma}{\beta}$ then \hat{x}_1 becomes unstable stable.

Solution to Exercise 2

(a)

```
function [t, X] = forward_euler(f, t0, tend, h, x0)
% FORWARD_EULER Implements the Forward Euler Method for solving
%               (non-linear discrete-time system)
%
% The Forward Euler update rule is:
%        $x_{k+1} = x_k + h * f(t_k, x_k)$ 

% number of steps
N = (tend - t0) / h;

% dimension of the state
n = length(x0);

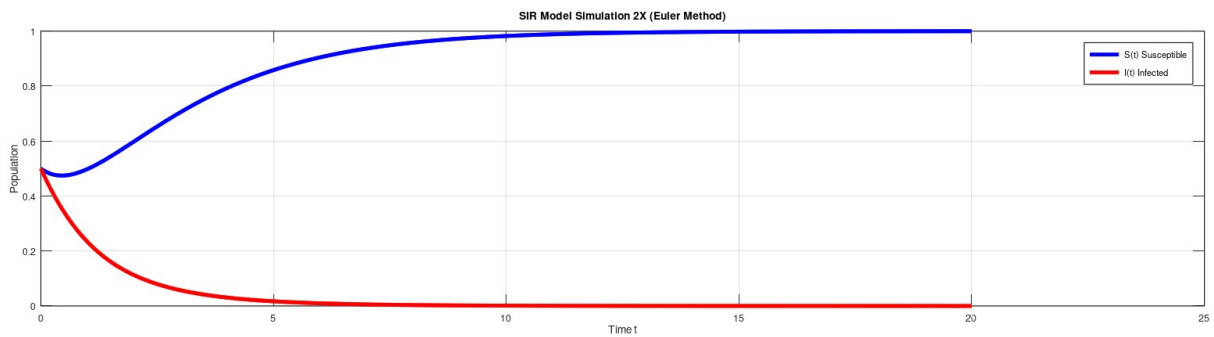
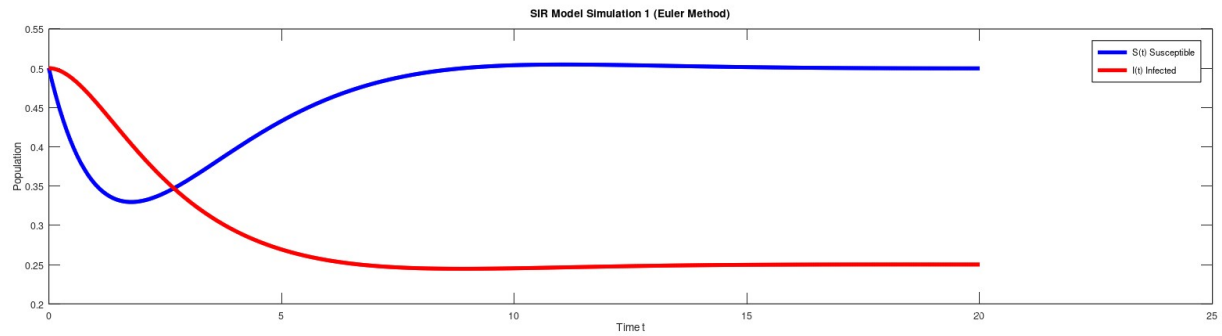
% give outputs empty values
t = zeros(1, N+1);
X = zeros(n, N+1);

% initial conditions
t(1) = t0;
X(:,1) = x0;

% Euler time-stepping loop
for k = 1:N
    X(:, k+1) = X(:,k) + h * f(t(k), X(:, k));
    t(k+1)    = t(k) + h;
end
```

Solution to Exercise 2 cont.

(b)



The First Simulation:

- $\beta = 1$
- $\gamma = 0.5$
- $\mu = 0.5$
- $P = 1$

$$\frac{\gamma}{\beta} = 0.5 < P = 1 > \frac{\gamma}{\beta}$$

The state in this simulation is converging to the second steady state of the simplified SIR model from Exercise 1.

$$\hat{x}_2 = \begin{bmatrix} \frac{\gamma}{\beta} \\ \frac{\mu(P - \frac{\gamma}{\beta})}{\gamma + \mu} \end{bmatrix}$$

Solution to Exercise 2 cont.

The Second Simulation:

- $\beta = 0.5$
- $\gamma = 1$
- $\mu = 0.5$
- $P = 1$

$$\frac{\gamma}{\beta} = 2 > P = P < \frac{\gamma}{\beta}$$

The state in this simulation is converging to the first steady of the simplified SIR model from Exercise 1.

$$\hat{x}_1 = \begin{bmatrix} P \\ 0 \end{bmatrix}$$

Theoretical results from Exercise 1:

If $P > \frac{\gamma}{\beta}$ then \hat{x}_1 is an unstable steady state and if $P < \frac{\gamma}{\beta}$ then \hat{x}_1 becomes asymptotically stable.

- In simulation 2: $P < \frac{\gamma}{\beta}$, and the method converges to \hat{x}_1 . i.e, \hat{x}_1 is an asymptotically stable steady state.

- In simulation 1: $P > \frac{\gamma}{\beta}$, and the method converges to \hat{x}_2 . i.e, \hat{x}_1 is an unstable steady state.

This confirms the results from Exercise 1:

If the overall population is small enough compared to the recovery rate over the infection rate, then the disease will naturally vanish without additional actions.

Solution to Exercise 3

(a)

With $k = 10$, $m=1$, and $d=0$ the linear mass-spring-damper system becomes:

$$\dot{x}(t) = \begin{bmatrix} 0 & 1 \\ -10 & 0 \end{bmatrix} x(t)$$

To analyze the stability of the system analytically, we first need eigenvalues:

$$0 = (0 - \lambda)(0 - \lambda) + 10 \quad (11)$$

$$= \lambda^2 + 10 \quad (12)$$

$$\lambda^2 = -10 \quad (13)$$

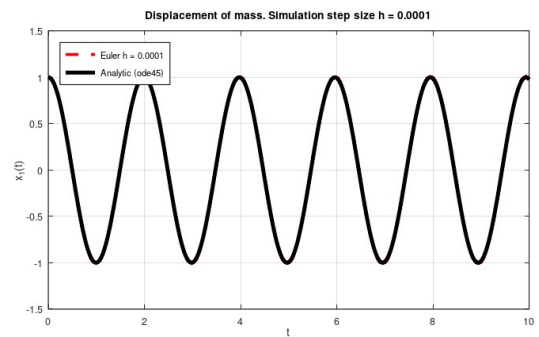
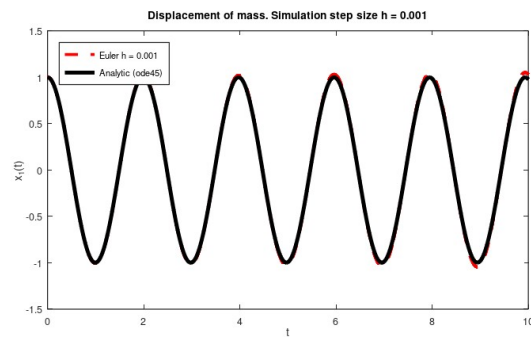
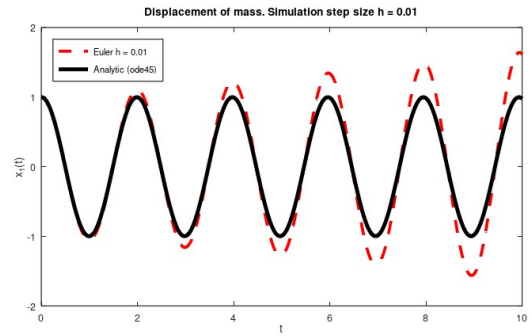
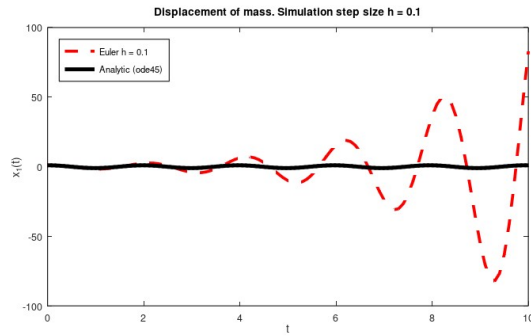
$$\lambda = \sqrt{-1}\sqrt{10} \quad (14)$$

$$\lambda = \pm i \sqrt{10} \quad (15)$$

The spectral abscissa(A) = 0, so the system is stable. I suspect the displacement and velocity to oscillate indefinitely with constant amplitude. This is comes from damping = 0.

Solution to Exercise 3 cont.

(b)



For the first three step sizes, error in the Euler method can be seen on the plots. However, for the step size $h = 0.0001$, you can not see any error.

Solution to Exercise 4

(a)

```
function [t, X] = heun(f, t0, tend, h, x0)
% HEUN'S METHOD Implements Heun's Method for solving dx/dt = f(t, x(t)).

% number of steps
N = round((tend - t0)/h);

% dimension of the state
n = length(x0);

% preallocate
t = zeros(1, N+1);
X = zeros(n, N+1);

% initial conditions
t(1) = t0;
X(:,1) = x0;

% Heun time stepping loop
for k = 1:N
    tk = t(k);
    xk = X(:,k);

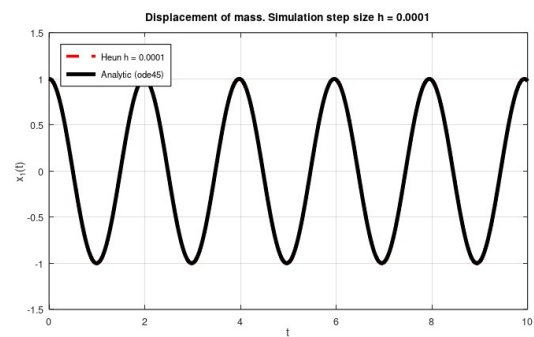
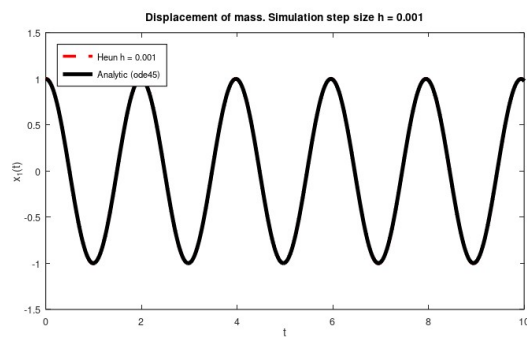
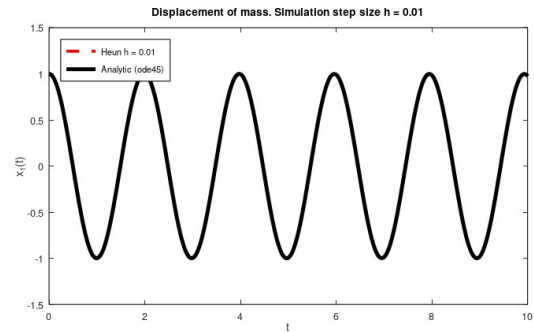
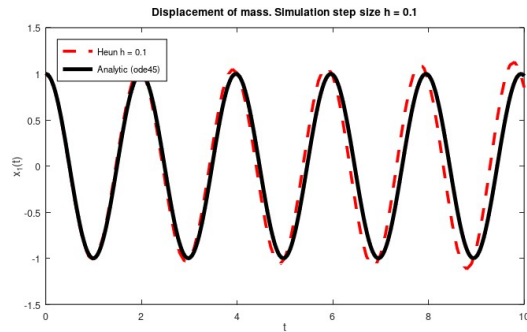
    % K1
    k1 = f(tk, xk);
    x_tilde = xk + h * k1;

    % K2
    k2 = f(tk + h, x_tilde);

    % update
    X(:,k+1) = xk + (h/2) * (k1 + k2);
    t(k+1) = tk + h;
end
end
```

Solution to Exercise 4 cont.

(b)



These results differ from the experiments with Euler's method as the error can no longer be seen in the step size $h = 0.01$. Also, the simulation at step size $h = 0.1$ is much closer to the true analytical solution than for Euler's method. This method does not overshoot the amplitude as much as Euler's method does in the first couple step sizes ($h=0.1$, $h=0.01$).

Bonus Exercise

$$\begin{aligned}w_{k+1} &= w_k + h * f(w_k) \\&= w_k + hAw_k \\&= (I + hA)w_k\end{aligned}$$

Substitute known values in:

$$w_{k+1} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + h \begin{bmatrix} -10 & 5 \\ 0 & -20 \end{bmatrix}$$

$$w_{k+1} = \begin{bmatrix} 1 - 10h & 5h \\ 0 & 1 - 20h \end{bmatrix} w_k$$

For the Forward Euler Method to be asymptotically stable, the transition matrix of the discretization must have a spectral radius less than 1, i.e., inside the unit circle:

$$\begin{aligned}|1 - 10h| &< 1 \\&= -1 < 1 - 10h < 1\end{aligned}$$

Starting with $-1 < 1 - 10h$:

$$\begin{aligned}-2 &< -10h \\h &< 0.2\end{aligned}$$

Now for $1 - 10h < 1$:

$$\begin{aligned}-10h &< 0 \\h &> 0\end{aligned}$$

Now looking at the other eigenvalue:

$$\begin{aligned}|1 - 20h| &< 1 \\-1 &< 1 - 20h < 1 \\-1 &< 1 - 20h, 1 - 20h < 1 \\-1 &< 1 - 20h \\-2 &< -20h \\h &< 0.1\end{aligned}$$

$$\begin{aligned}1 - 20h &< 1 \\-20h &< 0 \\h &> 0\end{aligned}$$

By taking the Union of these intervals, $0 < h < 0.1$