

Homework 3.

Amath 383

Introduction to Continuous Mathematical Modeling

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Due: 10/25/22 at 11:59pm to Gradescope

Directions:

Complete all exercises as neatly as possible. Up to 3 points may be deducted for homework that is illegible and/or poorly organized. You are encouraged to type homework solutions, and a half bonus point will be awarded to students who use \LaTeX . (Check out my \LaTeX beginner document and overleaf.com if you are new to \LaTeX .) If you prefer not to type homeworks, I ask that **homeworks be scanned.** (I will not accept physical copies.) In addition, **homeworks must be in .pdf format.**

Pro-Tips:

- You have access to the textbook, which inspired some of these exercises. You may find that the textbook offers alternative explanations that can help you solve these exercises.
- If a result was already derived in class, no need to rederive it here. Just make sure you cite where in the lecture notes the result your using can be found.
- Teamwork makes the dream work, but please indicate at the top of your assignment who your collaborators are.
- Don't wait until the last minute. ☺

Exercise 1. (Unit 1.5)

Consider the model equation

$$\frac{dP}{dt} = 1 - rP + P^2. \quad (1)$$

- a. For what value(s) of r does (1) have two fixed points? How about one fixed point? No fixed points?

Based on discriminate of quadratic equation is positive, negative or zero based on r .

$$\frac{-(-r) \pm \sqrt{(-r)^2 - 4(1)(1)}}{2(1)}$$

$$r^2 - 4 > 0?$$

Two fixed points when $r < -2, r > 2$,

One fixed points when $r = -2, r = 2$,

No fixed points when $-2 < r < 2$

- b. Show, using the appropriate conditions, that this model equation has two saddle-node bifurcations. What are the bifurcation points?

$r = 2$ the model equation has two saddle node points because the bifurcation points $(1, 2), (-1, -2)$ satisfies the following saddle point bifurcation equations:

$$F(P^*, r^*) = F(1, 2) = 1 - (2)(1) + (1)^2 = 1 - 2 + 1 = 0$$

$$F(P^*, r^*) = F(-1, -2) = 1 - (-2)(-1) + (-1)^2 = 1 - 2 + 1 = 0$$

$$\frac{dF}{dP}(P^*, r^*) = -r + 2P = -2 + 2 = 0$$

$$\frac{dF}{dP}(P^*, r^*) = -r + 2P = 2 - 2 = 0$$

$$\frac{dF}{dr}(P^*, r^*) = -P = -1 \neq 0$$

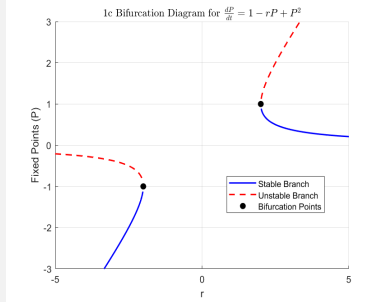
$$\frac{dF}{dr}(P^*, r^*) = -P = 1 \neq 0$$

$$\frac{d^2F}{dP^2}(P^*, r^*) = 2 \neq 0$$

At $r = 2$, the quadratic equation $P^2 - rP + 1 = 0$ has a double root, meaning the two fixed points merge. This is characteristic of saddle-node bifurcations.

- c. Using software of your choice, plot a bifurcation diagram for $-5 \leq r \leq 5$. Be sure to identify the stable and unstable branches in your plot.

Matlab is the software of my choice.



Now consider the model equation

$$\frac{dP}{dt} = P(r - e^P). \quad (2)$$

- d. For what value(s) of r does (2) have two fixed points? How about one fixed point? No fixed points?

$$\frac{dP}{dt} = P(r - e^P) = 0$$

Two fixed points when $r > 0, 0 < r < 1$, at $P = 0, P = \ln(r)$ because natural log is only defined for $r > 0$

One fixed points when $r \leq 0$ and $r = 1$ because there is always at least one fixed point at $P = 0$ and when $r = 1$ then $P = \ln(1) = 0$ so then $P = 0$ is the only fixed point.

- e. Show, using the appropriate conditions, that this model equation has a transcritical bifurcation. What is the bifurcation point?

Transcritical bifurcation

The bifurcation point is $(0, 1)$

$$F(P^*, r^*) = 0(1 - e^0) = 0$$

$$\frac{dF}{dP}(P^*, r^*) = (1 - e^{P^*}) = (1 - e^0) = 1 - 1 = 0$$

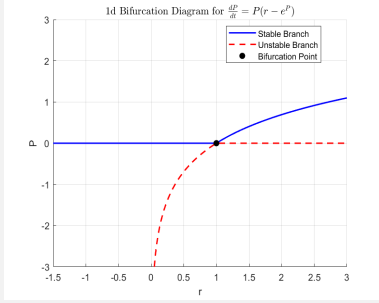
$$\frac{dF}{dr}(P^*, r^*) = P = 0$$

$$\frac{d^2F}{dPdr}(P^*, r^*) = \frac{dF}{dr}(r - e^P) = 1 \neq 0$$

$$\frac{d^2F}{dP^2}(P^*, r^*) = -e^P = -e^0 = -1 \neq 0$$

- f. Using software of your choice, plot a bifurcation diagram for $-1.5 \leq r \leq 3$. Be sure to identify the stable and unstable branches in your plot.

Matlab is the software of my choice



Exercise 2. (Units 1.5-1.6)

The FitzHugh-Nagumo equations were proposed independently by Richard FitzHugh in 1961 and Jin-ichi Nagumo in 1962 as a model for the firing of individual nerve cells, such as neurons in the brain. The (dimensionless) model equations are as follows:

$$\frac{dV}{dt} = V - \frac{1}{3}V^3 - W + I, \quad (3a)$$

$$\frac{dW}{dt} = \frac{1}{T_R} \left(V - \frac{W}{T_F} \right), \quad (3b)$$

where V represents the voltage potential across the membrane of the cell, I represents the net current due to ions flowing into and out of the cell, W is an artificial relaxation variable that controls how V relaxes after the cell fires, $T_R > 0$ is a constant that controls the time scale of the relaxation period between cell firings, and $T_F > 0$ is a constant that controls the time scale for the decay of the relaxation variable, allowing the cell to unrelax and refire.

- a. Suppose $T_R \gg 1$, meaning $\frac{dW}{dt}$ is effectively zero. In this case, what is W as a function of V and T_F ?

$$\begin{aligned} \frac{dW}{dt} &= \frac{1}{T_R} \left(V - \frac{W}{T_F} \right) = 0 \\ \frac{1}{T_R} \left(V - \frac{W}{T_F} \right) &= 0 \\ \left(V - \frac{W}{T_F} \right) &= 0 \\ V &= \frac{W}{T_F} \\ \boxed{W} &= \boxed{VT_F} \end{aligned}$$

- b. Show that, when $T_R \gg 1$ and there is no net current flowing across the cell membrane, we have

$$\frac{dV}{dt} = V \left(1 - T_F - \frac{1}{3}V^2 \right). \quad (4)$$

Net current = 0 so $I = 0$

$$\frac{dV}{dt} = V - \frac{1}{3}V^3 - W + I$$

$$\frac{dV}{dt} = V - \frac{1}{3}V^3 - W + (0)$$

$$\frac{dV}{dt} = V\left(1 - \frac{1}{3}V^2 - \frac{W}{V}\right)$$

$$\frac{W}{V} = T_F$$

$$\frac{dV}{dt} = V\left(1 - \frac{1}{3}V^2 - T_F\right)$$

- c. Show, using the appropriate conditions, that (4) has a Pitchfork bifurcation with respect to the parameter T_F . Sketch the bifurcation diagram. Is this Pitchfork bifurcation sub- or super-critical?

$$\frac{dV}{dt} = V\left(1 - T_F - \frac{1}{3}V^2\right) = 0$$

$V = 0$ obvious

$$1 - T_F - \frac{1}{3}V^2 = 0$$

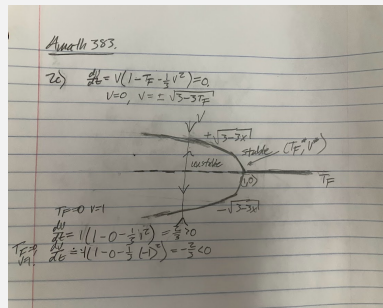
$$-\frac{1}{3}V^2 = -1 + T_F$$

$$V^2 = 3 - 3T_F$$

$$V = \pm\sqrt{3 - 3T_F}$$

Fixed Points at $V = 0$, $V = +\sqrt{3 - 3T_F}$, $V = -\sqrt{3 - 3T_F}$

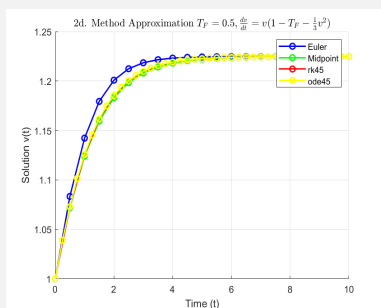
Supercritical bifurcation.



- d. Suppose initially $V(0) = 1$. Solve (4) with $T_F = 0.5$ by the following numerical methods:

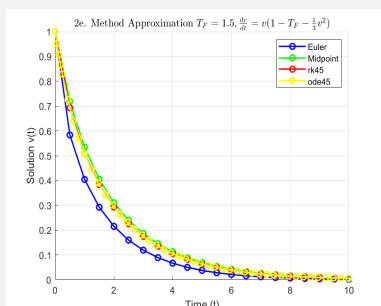
- Euler's method,
- The midpoint method,
- The Runge-Kutta method,
- MATLAB's ode45 method.

Use $N = 20$ time steps over the interval $0 \leq t \leq 10$. The code for these methods can be found on the Homework 3 Canvas page. Superimpose the graphs of each numerical solution onto one graph and attach this graph to your homework assignment. *Include a legend on your graph to distinguish each of the numerical solutions.*



- e. Repeat the analysis in (d) but with $T_F = 1.5$. How are your plots qualitatively different than those in (d)? How can your analysis in (c) account for this difference?

Remark: As you have seen, T_F must be sufficiently small in order for the cell to unrelax and build up voltage potential. (Otherwise, the cell continues to relax.) This makes sense, since a small T_F means that W decays quickly, allowing V to increase more rapidly.



How are your plots qualitatively different than those in (d)?

In part (c) the plots increase from the initial condition 1 to a stable equilibrium of about 1.23

In part (d) the plots decrease from the initial condition 1 to a stable equilibrium of 0

How can your analysis in (c) account for this difference?

In part (c) the bifurcation diagram shows that when $T_F < 1$ there are three equilibrium two stable with values $V = \pm\sqrt{3-3T_F}$ and one unstable with value $V = 0$. An initial condition of 1 would go towards the $+\sqrt{3-3T_F}$ in the long term. While when $T_F > 1$ there is only one stable equilibrium at $V = 0$. Meaning an initial condition of 1 would go to 0 in the long term.

Exercise 3. (Unit 1.6)

Consider the logistic equation with initial growth rate $r = 1$ and carrying capacity $K = 1$:

$$\frac{dP}{dt} = P(1 - P).$$

- a. Find the exact solution of this equation satisfying the initial condition $P(0) = \frac{1}{2}$.

$$\begin{aligned}\frac{dP}{dt} &= P(1 - P). \\ \frac{dP}{P(1-P)} &= dt \\ \ln(P) - \ln(P-1) &= t + C_1 \\ \ln\left(\frac{P}{P-1}\right) &= t + C_1 \\ \frac{P}{P-1} &= e^{t+C_1} \\ P &= \frac{1}{1+C_1 e^{-t}}\end{aligned}$$

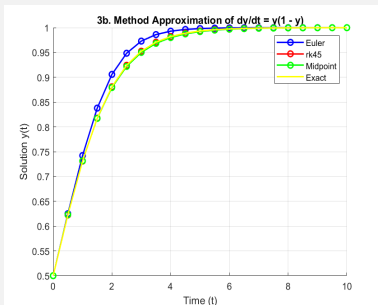
$$\begin{aligned}P(0) &= \frac{1}{2} = \frac{1}{1+C_2 e^{-(0)}} \\ &= \frac{1}{2} = \frac{1}{1+C_2} \\ C_2 &= 1\end{aligned}$$

$$P(t) = \frac{1}{1 + e^{-t}}$$

b. Find the solution of this equation satisfying the initial condition $P(0) = 1/2$ by the following numerical methods:

- Euler's method,
- The midpoint method,
- The Runge-Kutta method.

Use $N = 20$ time steps over the interval $0 \leq t \leq 10$. Superimpose the graph of the exact solution and the graphs of each of the numerical solutions onto one graph and attach your graph to your homework assignment. *Include a legend on your graph to distinguish each of the solutions.*



Define the error of each method as the maximum difference in absolute value between the exact solution and the numerical solution over $0 \leq t \leq 10$:

$$\mathcal{E} = \max_{0 \leq t \leq 10} |P_{\text{exact}}(t) - P_{\text{numerical}}(t)|.$$

What is the error of each numerical method?

Hint: You may find MATLAB's built-in max function useful.

$error_{eu} = 0.0250$
 $error_{mi} = 0.0024$
 $error_{rk} = 3.6566e - 05$

- c. Obtain the errors for each numerical method using $N = 200$ time steps. Have the errors improved?

$error_{eu} = 0.0023$
 $error_{mi} = 1.8708e - 05$
 $error_{rk} = 2.9289e - 09$
 Error values have decreased for all of them.

- d. According to theory, the error \mathcal{E} of a finite-difference method is approximately given by

$$\mathcal{E} = Ch^p,$$

where h is the step size used for the method, p is the order of the method, and C is a constant of proportionality. Given two errors \mathcal{E}_1 and \mathcal{E}_2 for two different step sizes h_1 and h_2 , show that

$$p = \frac{\ln\left(\frac{\mathcal{E}_2}{\mathcal{E}_1}\right)}{\ln\left(\frac{h_2}{h_1}\right)}.$$

$$\begin{aligned}
 \mathcal{E}_1 &= Ch_1^p, \mathcal{E}_2 = Ch_2^p \\
 \ln(\mathcal{E}_1) &= p \ln(Ch_1), \ln(\mathcal{E}_2) = p \ln(Ch_2) \\
 \ln(\mathcal{E}_2) - \ln(\mathcal{E}_1) &= p \ln(Ch_2) - p \ln(Ch_1) \\
 \ln(\mathcal{E}_2) - \ln(\mathcal{E}_1) &= p(\ln(Ch_2) - \ln(Ch_1)) \\
 \ln\left(\frac{\mathcal{E}_2}{\mathcal{E}_1}\right) &= p \ln\left(\frac{Ch_2}{Ch_1}\right) \\
 \ln\left(\frac{\mathcal{E}_2}{\mathcal{E}_1}\right) &= p \ln\left(\frac{Ch_2}{Ch_1}\right) \\
 \ln\left(\frac{\mathcal{E}_2}{\mathcal{E}_1}\right) &= p \ln\left(\frac{h_2}{h_1}\right) \\
 p &= \frac{\ln\left(\frac{\mathcal{E}_2}{\mathcal{E}_1}\right)}{\ln\left(\frac{h_2}{h_1}\right)}
 \end{aligned}$$

- e. Using the errors computed in (b) and (c) and the formula derived in (d), estimate the order of Euler's method, the midpoint method, and the Runge-Kutta method. Do these orders match what you expect?

$$p = \frac{\ln\left(\frac{\varepsilon}{\varepsilon_1}\right)}{\ln\left(\frac{h_2}{h_1}\right)}$$

$$p_{eu} = \frac{\ln\left(\frac{0.0023}{0.0250}\right)}{\ln\left(\frac{200}{20}\right)} = 1.0362 \approx 1$$

$$p_{mi} = \frac{\ln\left(\frac{1.8708e-05}{0.0024}\right)}{\ln\left(\frac{200}{20}\right)} = 2.1082 \approx 2$$

$$p_{rk} = \frac{\ln\left(\frac{2.9289e-09}{3.6566e-05}\right)}{\ln\left(\frac{200}{20}\right)} = 4.0964 \approx 4$$

Yes the orders match what I expected with higher the order the lower the error.

euler method is first order

midpoint method is second order

fourth order runge-kutta method is indeed fourth order

```

1 %% AMATH 383 HW 3
2
3 %% Excercise 1c
4 %dp/dt = 1 - rP + P^2
5
6 % Define the range for r and the function for the fixed points
7 r_values = linspace(-5, 5, 500);
8 fixed_points_positive = NaN(length(r_values), 1);
9 fixed_points_negative = NaN(length(r_values), 1);
10
11 % Compute fixed points based on the discriminant condition
12 for i = 1:length(r_values)
13     r = r_values(i);
14     discriminant = r^2 - 4;
15     if discriminant >= 0 % Real roots exist
16         % quadformula for P.
17         root1 = (r + sqrt(discriminant)) / 2;
18         root2 = (r - sqrt(discriminant)) / 2;
19         fixed_points_positive(i) = root2;
20         fixed_points_negative(i) = root1;
21     end
22 end
23
24 % Plot the bifurcation diagram
25 figure;
26 hold on;
27 plot(r_values, fixed_points_positive, 'b-', 'LineWidth', 1.5); %
    Stable Branch
28 plot(r_values, fixed_points_negative, 'r--', 'LineWidth', 1.5);
    % Unstable Branch
29
30 % Highlight bifurcation points
31 plot([-2, 2], [-1, 1], 'ko', 'MarkerFaceColor', 'k', 'MarkerSize
    ', 6); % Bifurcation Points
32
33 % Formatting the plot
34 ax = gca; % gridlines and axis features.
35 ax.XLim = [-5, 5];
36 ax.YLim = [-3, 3];
37 ax.XGrid = 'on';
38 ax.YGrid = 'on';
39 title('1c Bifurcation Diagram for  $\frac{dP}{dt} = 1 - rP + P^2$ 
    ', 'Interpreter', 'latex');
40 xlabel('r');
41 ylabel('Fixed Points (P)');
42 legend('Stable Branch', 'Unstable Branch', 'Bifurcation Points',

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        'Location', 'best');
43 hold off;
44
45 %% Excercise 1d
46 %dP/dt = P(r-e^P)
47
48 % Define the range for r and the function for the fixed points
49 r_values = linspace(-1.5, 3, 500);
50 fixed_points_positive = NaN(length(r_values), 1);
51 fixed_points_negative = NaN(length(r_values), 1);
52
53 % Compute fixed points based on the discriminant condition
54 for i = 1:length(r_values)
55     r = r_values(i);
56     if r < 1
57         fixed_points_positive(i) = 0;
58     end
59     if r > 0 && r < 1
60         fixed_points_negative(i) = log(r);
61     end
62     if r > 1
63         fixed_points_positive(i) = log(r);
64         fixed_points_negative(i) = 0;
65     end
66 end
67
68 % Plot the bifurcation diagram
69 figure;
70 hold on;
71 plot(r_values, fixed_points_positive, 'b-', 'LineWidth', 1.5); %
    Stable Branch
72 plot(r_values, fixed_points_negative, 'r--', 'LineWidth', 1.5);
    % Unstable Branch
73
74 % Highlight bifurcation points
75 plot([1], [0], 'ko', 'MarkerFaceColor', 'k', 'MarkerSize', 6); %
    Bifurcation Points
76
77 % Formatting the plot
78 ax = gca; % gridlines and axis features.
79 ax.XLim = [-1.5, 3];
80 ax.YLim = [-3, 3];
81 ax.XGrid = 'on';
82 ax.YGrid = 'on';
83 title('1d Bifurcation Diagram for  $\frac{dP}{dt} = P(r-e^P)$ ',
    'Interpreter', 'latex');
84 xlabel('r');

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85 ylabel('P');
86 legend('Stable Branch', 'Unstable Branch', 'Bifurcation Point',
      'Location', 'best');
87 hold off;
88
89
90
91 %% Excercise 2d
92 V_0 = 1;
93 T_F = 0.5
94 ode_RHS = @(t,v) v*(1 - T_F - (1/3)*v^2);
95
96 [t2d_eu, soln2d_eu] = UW_euler_method_383(ode_RHS, [0, 10], 20,
      V_0)
97 [t2d_mi, soln2d_mi] = UW_midpoint_method_383(ode_RHS, [0, 10] ,
      20, V_0)
98 [t2d_rk, soln2d_rk] = UW_rk4_method_383(ode_RHS, [0, 10], 20,
      V_0)
99 [t2d_45, soln2d_45] = ode45(ode_RHS,[0,10],V_0)
100
101 figure;
102 hold on
103 ax = gca; % gridlines and axis features.
104 %ax.XLim = [-1.5, 3];
105 ax.YLim = [0, 2.0];
106 ax.XGrid = 'on';
107 ax.YGrid = 'on';
108
109 plot(t2d_eu, soln2d_eu, '-o', 'LineWidth', 1.5, 'Color', 'blue');
110 plot(t2d_mi, soln2d_mi, '-o', 'LineWidth', 1.5, 'Color', 'green');
111 plot(t2d_rk, soln2d_rk, '-o', 'LineWidth', 1.5, 'Color', 'red');
112 plot(t2d_45, soln2d_45, '-o', 'LineWidth', 1.5, 'Color', 'yellow')
113 ;
114 xlabel('Time (t)');
115 ylabel('Solution v(t)');
116
117 title('2d. Method Approximation $T_F = 0.5, \frac{dv}{dt} = v(1$
      - $T_F - \frac{1}{3}v^2)$', 'Interpreter', 'latex');
118 legend('Euler', 'Midpoint', 'rk45', 'ode45')
119 grid on;
120 hold off
121
122 %% Excercise 2e
123 T_F = 1.5
124 ode_RHS = @(t,v) v*(1 - T_F - (1/3)*v^2);
125 [t2e_eu, soln2e_eu] = UW_euler_method_383(ode_RHS, [0, 10], 20,

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1)
126 [t2e_mi, soln2e_mi] = UW_midpoint_method_383(ode_RHS, [0, 10] ,
    20, 1)
127 [t2e_rk, soln2e_rk] = UW_rk4_method_383(ode_RHS, [0, 10], 20, 1)
128 [t2e_45, soln2e_45] = ode45(ode_RHS,[0,10],1)
129
130 figure;
131 hold on
132 ax = gca; % gridlines and axis features.
133 %ax.XLim = [-1.5, 3];
134 ax.YLim = [0, 2.0];
135 ax.XGrid = 'on';
136 ax.YGrid = 'on';
137
138 plot(t2e_eu, soln2e_eu, '-o', 'LineWidth', 1.5, 'Color', 'blue');
139 plot(t2e_mi, soln2e_mi, '-o', 'LineWidth', 1.5, 'Color', 'green');
140 plot(t2e_rk, soln2e_rk, '-o', 'LineWidth', 1.5, 'Color', 'red');
141 plot(t2e_45, soln2e_45, '-o', 'LineWidth', 1.5, 'Color', 'yellow')
    ;
142 xlabel('Time (t)');
143 ylabel('Solution v(t)');
144 title('2e. Method Approximation $T_F = 1.5, \frac{dv}{dt} = v(1$
    - $T_F - \frac{1}{3}v^2)$', 'Interpreter', 'latex');
145 legend('Euler', 'Midpoint', 'rk45', 'ode45')
146 grid on;
147 hold off
148
149 %% Excercise 3.
150 %num_steps = 20;
151 %init_cond = 0.5;
152 %t_start = 0;
153 %t_end = 10;
154
155 num_steps = 20;
156 t_range = [0,10];
157 h = (t_range(2)-t_range(1))/num_steps;
158 t = t_range(1):h:t_range(2);
159 y_exact = exp(t) ./ (1 + exp(t));
160
161 ode_RHS = @(t,y) y*(1-y);
162 [t3b_eu, soln3b_eu] = UW_euler_method_383(ode_RHS, [0, 10], 20,
    0.5)
163 [t3b_rk, soln3b_rk] = UW_rk4_method_383(ode_RHS, [0, 10], 20,
    0.5)
164 [t3b_mi, soln3b_mi] = UW_midpoint_method_383(ode_RHS, [0, 10] ,
    20, 0.5)
165

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166 % Plot the solution
167 figure;
168 hold on
169 plot(t3b_eu, soln3b_eu, '-o', 'LineWidth', 1.5, 'Color', 'blue');
170 plot(t3b_rk, soln3b_rk, '-o', 'LineWidth', 1.5, 'Color', 'red');
171 plot(t3b_mi, soln3b_mi, '-o', 'LineWidth', 1.5, 'Color', 'green');
172 plot(t, y_exact, '-', 'LineWidth', 1.5, 'Color', 'yellow' );
173 xlabel('Time (t)');
174 ylabel('Solution y(t)');
175 title('3b. Method Approximation of  $dy/dt = y(1 - y)$ ');
176 legend('Euler', 'rk45', 'Midpoint', 'Exact')
177 grid on;
178 hold off
179
180 %% Part b error N = 20
181 num_steps = 20;
182 t_range = [0,10];
183 h = (t_range(2)-t_range(1))/num_steps;
184 t = t_range(1):h:t_range(2);
185 y_exact = exp(t) ./ (1 + exp(t));
186
187 % Compute error
188 error_eu = max(abs(y_exact - soln3b_eu))
189 error_rk = max(abs(y_exact - soln3b_rk))
190 error_mi = max(abs(y_exact - soln3b_mi))
191
192
193 %% Part c error N = 200
194 num_steps = 200;
195 t_range = [0,10];
196 h = (t_range(2)-t_range(1))/num_steps;
197 t = t_range(1):h:t_range(2);
198 y_exact = exp(t) ./ (1 + exp(t));
199
200 [t3c_eu, soln3c_eu] = UW_euler_method_383(ode_RHS, [0, 10], 200,
    0.5);
201 [t3c_rk, soln3c_rk] = UW_rk4_method_383(ode_RHS, [0, 10], 200,
    0.5);
202 [t3c_mi, soln3c_mi] = UW_midpoint_method_383(ode_RHS, [0, 10] ,
    200, 0.5);
203
204 % Compute error
205 error_eu = max(abs(y_exact - soln3c_eu))
206 error_rk = max(abs(y_exact - soln3c_rk))
207 error_mi = max(abs(y_exact - soln3c_mi))
208
209

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210
211 %%
212
213 function [t,soln] = UW_euler_method_383(ode_RHS, t_range,
    num_steps,init_cond)
214     h = (t_range(2)-t_range(1))/num_steps;
215     t = t_range(1):h:t_range(2);
216     soln = [init_cond,nan(1,length(t)-1)];
217 % ode_RHS is the function handle of the RHS of the ODE, i.e.,
218 % num_steps is the number of time steps from t_start
219 % init_cond is the initial condition
220 %t_range = [t_start t_end];
221
222     for j = 1:length(t)-1
223         soln(j+1) = soln(j) + h*ode_RHS(t(j),soln(j));
224     end
225 end
226
227
228 function [t,soln] = UW_rk4_method_383(ode_RHS,t_range,num_steps,
    init_cond)
229
230 % ode_RHS is the function handle of the RHS of the ODE, i.e.,
231 %ode_RHS = @(t,y) y
232 % t_range = [t_start t_end];
233 % num_steps is the number of time steps from t_start
234 % init_cond is the initial condition
235
236 h = (t_range(2)-t_range(1))/num_steps;
237 t = t_range(1):h:t_range(2);
238 soln = [init_cond,nan(1,length(t)-1)];
239
240 for j = 1:length(t)-1
241     K1 = ode_RHS(t(j),soln(j));
242     K2 = ode_RHS(t(j)+h/2,soln(j)+h*K1/2);
243     K3 = ode_RHS(t(j)+h/2,soln(j)+h*K2/2);
244     K4 = ode_RHS(t(j)+h,soln(j)+h*K3);
245     soln(j+1) = soln(j) + h/6*(K1+2*K2+2*K3+K4);
246 end
247 end
248
249
250 function [t,soln] = UW_midpoint_method_383(ode_RHS,t_range,
    num_steps,init_cond)
251
252 % ode_RHS is the function handle of the RHS of the ODE, i.e.,
253 % ode_RHS = @(t,y) [insert RHS of ODE here]

```

```

254 % t_range = [t_start t_end]
255 % num_steps is the number of time steps from t_start
256 % init_cond is the initial condition
257
258 h = (t_range(2)-t_range(1))/num_steps;
259 t = t_range(1):h:t_range(2);
260 soln = [init_cond,nan(1,length(t)-1)];
261
262 for j = 1:length(t)-1
263     soln(j+1) = soln(j) + h*ode_RHS(t(j)+h/2,soln(j)+h/2*...
264         ode_RHS(t(j),soln(j)));
265 end
266 end

```