

MECH 6313 - Homework 3

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1 Problem 1

Problem: Let

$$\begin{aligned}\dot{x}_1 &= -x_1 + x_2 \\ \dot{x}_2 &= \frac{x_1^2}{1+x_1^2} - 0.5x_2\end{aligned}\tag{1}$$

Define an shifted system, linerize that system, and find the center manifold to analyze the stability properties. Then use numerical simulation to plot the phase portrait of the original coordinates and superimpose the shifted center manifold.

Solution:

1.1 Part a

Let a shifted set of state variables be defined as $\bar{x}_1 = x_1 - 1$ and $\bar{x}_2 = x_2 - 1$. The state variable equation can then be rewritten as

$$\begin{aligned}\dot{\bar{x}}_1 &= -\bar{x}_1 + \bar{x}_2 \\ \dot{\bar{x}}_2 &= \frac{(\bar{x}_1 + 1)^2}{(1 + \bar{x}_1^2)^2} - \frac{\bar{x}_2 + 1}{2}\end{aligned}\tag{2}$$

1.2 Part b

This system can then be linearized about the origin, resulting in the system matrix

$$A = \begin{bmatrix} -1 & 1 \\ \frac{1}{2} & -\frac{1}{2} \end{bmatrix}\tag{3}$$

whose eigenvalues are calculated as $\lambda_{1,2} = 0, -\frac{3}{2}$.

A transformation matrix

$$T = \begin{bmatrix} 1 & -2 \\ 1 & 1 \end{bmatrix}$$

can then be constructed with the associated eigenvectors to covert using

$$\begin{bmatrix} y \\ z \end{bmatrix} = T^{-1}x$$

to transform into the diagonalized system

$$\begin{aligned}\dot{y} &= A_1 y + g_1(y, z) \\ \dot{z} &= A_2 z + g_2(y, z)\end{aligned}\tag{4}$$

where $A_1 = 0$, $A_2 = -\frac{3}{2}$, $g_1(y, z) = 3z$, and $g_2(y, z) = \frac{(y - 2z + 1)^2}{(y - 2z + 1)^2 + 1} + \frac{2z - y + 1}{2}$.

An invariant manifold can then be defined as

$$\omega = z - h(y)$$

with $\dot{\omega}$ calculated as

$$\dot{\omega} = \dot{z} - \frac{\partial h}{\partial y} \dot{y}$$

To satisfy invariance, $z = h(y)$, which implies $\dot{\omega} = \omega = 0$. This implies that for an invariant manifold to exist the following must be true:

$$\dot{\omega} = 0 = A_2 h(y) + g_2(y, h(y)) - \frac{\partial h}{\partial y} [A_1 y + g_1(y, h(y))]\tag{5}$$

For a simple stability test on the invariant manifold, a Taylor series approximation of $h(y)$ can be used assuming that $h(0) = \left. \frac{dh}{dy} \right|_0 = 0$ around the origin:

$$h(y) = h_2 y^2 + O(y^3)\tag{6}$$

which would result in

$$\frac{dh}{dy} = 2h_2 y + O(y^2)\tag{7}$$

and the the following must hold:

$$\dot{\omega} = 0 = A_2 h(y) + g_2(y, (h_2 y^2 + O(y^3))) - (2h_2 y + O(y^2)) [A_1 y + g_1(y, (h_2 y^2 + O(y^3)))]\tag{8}$$

which can be manipulated to solve for h_2 given that

$$h_2 y^2 = \frac{1}{3} y^2$$

therefore

$$h_2 = \frac{1}{3}\tag{9}$$

1.3 Part c

At the equilibrium point $\bar{x} = (0,0)$, or $x = (1,1)$, the invariant manifold characterized with

$$\dot{\omega} = \dot{z} - \frac{dh}{dy}\dot{y} \quad (10)$$

$$= \dot{z} - 2/3y\dot{y} \quad (11)$$

thus it is a stable equilibrium point.

1.4 Part d

The phase portrait shown in Figure1 demonstrates the expected behavior. There is a stable equilibrium point where $\bar{x} = (0,0)$.

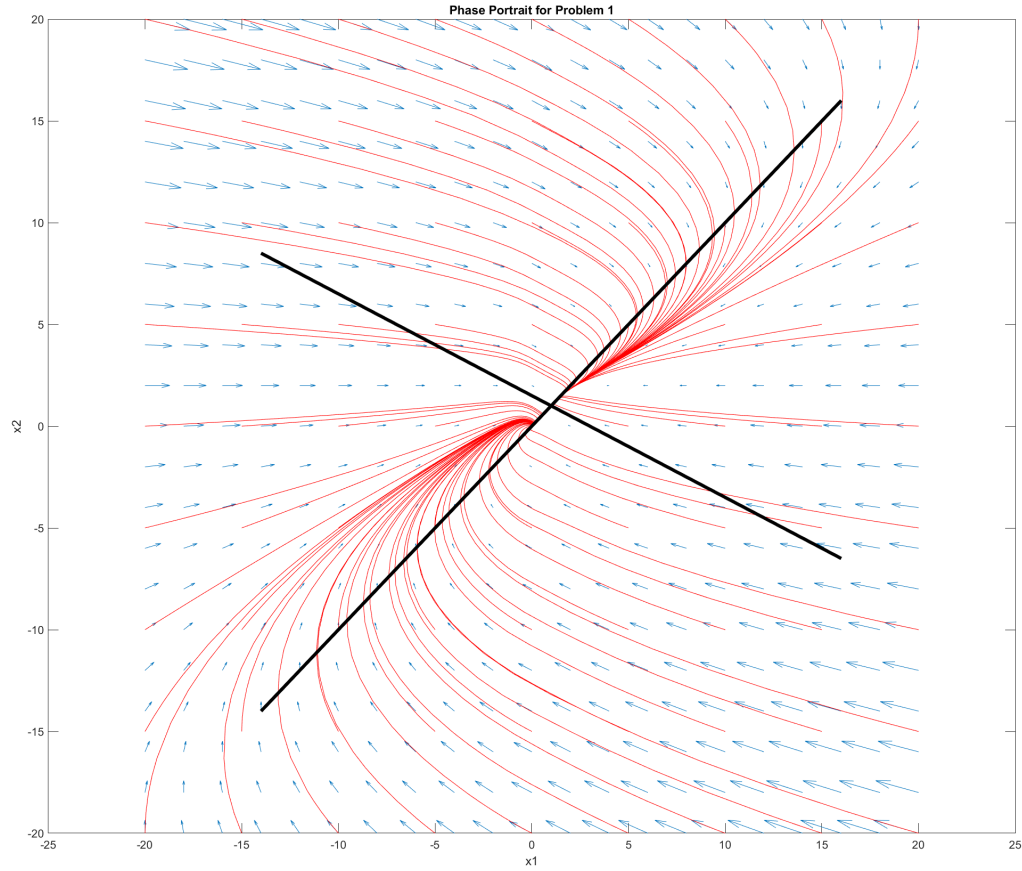


Figure 1: Phase Portrait for the original system.

2 Problem 2 - S 3.7.3

Problem: A simple model of a fishery is given as

$$\dot{N} = rN(1 - \frac{N}{K}) - H \quad (12)$$

where N represents the fish population, $H > 0$ is the number of fish harvested at a constant rate, and both r and K are constants.

Redefine the model in terms of x , τ , and h . Then plot the vector field for various values of h . Then identify h_c and classify and discuss the bifurcation.

Solution:

2.1 Part a

Let $x = N/K$, this can then be substituted as

$$\dot{N} = \frac{dN}{dt} = \frac{dx}{dt} = r(Kx)(1 - x) - H \quad (13)$$

$$\frac{1}{rK} \frac{dx}{dt} = x(1 - x) - \frac{H}{rK} \quad (14)$$

Let $h = \frac{H}{rK}$ and $\tau = rKt$,

$$\frac{dx}{d\tau} = x(1 - x) - h \quad (15)$$

2.2 Part b

The MATLAB code in AppendixA plots a vector field of the simplified fishery model as seen in Figure4.

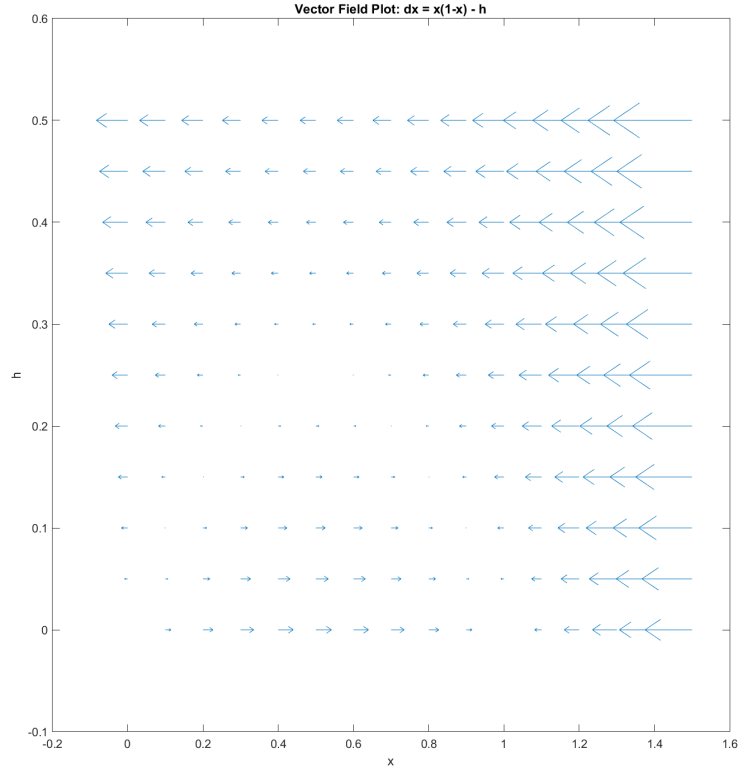


Figure 2: Vector field of the simple fishery model.

2.3 Part c

As is evident by observing the vector fields shown in 2 there exists an $h_c = 0.25$ where the bifurcation occurs. This is a form of fold bifurcation.

2.4 Part d

The long time behavior of this model results in either a stable steady-state population of fish or the extinction of the fish. When $h < h_c$, there is a point where the fishing will balance out the reproduction of the fish, but if the original population isn't large enough they could also become extinct. When $h > h_c$ there is an issue of overfishing and the fish will always become extinct.

3 Problem 3 - S 3.7.4

Problem: An improved model of a fishery is given as

$$\dot{N} = rN\left(1 - \frac{N}{K}\right) - H\frac{N}{A + N} \quad (16)$$

where N represents the fish population, $H > 0$ is the number of fish harvested at a constant rate, and both r , K , A are constants.

Define the biological interpretation of the parameter A . Redefine the model in terms of x , τ , and h . Find and analyze various fixed points depending on the values of a and h . Then analyze the bifurcation that occurs when $h = a$. Then find and classify the other bifurcation that occurs at $h = \frac{1}{4}(a + 1)^2$ for $a < a_c$. Finally plot the stability diagram for the system for (a, h) .

Solution:

3.1 Part a

When a population of fish is being fished, there is a portion of fish that are not possible to catch (such as eggs or fish that are too old).

3.2 Part b

Let $x = N/K$, this can then be substituted as

$$\dot{N} = \frac{dN}{dt} = \frac{dx}{dt} = r(Kx)(1 - x) - H\frac{Kx}{A + Kx} \quad (17)$$

$$\frac{1}{rK} \frac{dx}{dt} = x(1 - x) - \frac{H}{rK} \frac{Kx}{A + Kx} \quad (18)$$

$$= x(1 - x) - \frac{H}{rK} \frac{Kx}{K\left(\frac{A}{K} + x\right)} \quad (19)$$

$$= x(1 - x) - \frac{H}{rK} \frac{x}{\frac{A}{K} + x} \quad (20)$$

Let $h = \frac{H}{rK}$, $\tau = rKt$, and $a = \frac{A}{K}$

$$\frac{dx}{d\tau} = x(1 - x) - h\frac{x}{a + x} \quad (21)$$

3.3 Part c

The various fixed points of different regions can be determined by solving for when

$$\frac{dx}{d\tau} = 0$$

which can be found as the solution of

$$\frac{dx}{d\tau} = 0 = x(1 - x) - h \frac{x}{a + x} \quad (22)$$

$$\frac{hx}{a + x} = x(1 - x) \quad (23)$$

$$0 = x^3 + (a - 1)x^2 + (h - a)x \quad (24)$$

$$= x(x^2 + (a - 1)x + (h - a)) \quad (25)$$

which then can be solved using the quadratic formula such that the roots are given as

$$x = \left\{ 0, \frac{1 - a}{2} \pm \sqrt{\frac{a^2 + 2a - 4h + 1}{4}} \right\} \quad (26)$$

This solution indicates that there is a possibility of 1, 2, or 3 roots depending on the quantity:

$$a^2 + 2a - 4h + 1$$

When positive there are 3 equilibrium points, if equal to zero there is two, and if negative there is only one.

3.4 Part d

When looking exclusively near the equilibrium point at zero, a single bifurcation is evident when $h = a$. This is because the roots are found as

$$0 = x(x^2 + (a - 1)x + (0)) \quad (27)$$

$$0 = x^2(x + a - 1) \quad (28)$$

which means that there is now a double root at zero and the marginal stability is lost at that point. This is also the boundary for trans-critical bifurcation.

3.5 Part e

The other bifurcation shift point occurs when

$$0 = a^2 + 2a - 4h + 1 \quad (29)$$

$$4h = a^2 + 2a + 1 \quad (30)$$

$$h = \frac{1}{4}(a + 1)^2 \quad (31)$$

where a pitchfork bifurcation occurs.

3.6 Part f

The stability diagram shown in Figure3 shows 3 distinct regions, and 3 boundaries, that display different response characteristics. In the blue region there is a single real root occurring at the origin, which is the case where there is over fishing and the only stable state is extinction. In the yellow region there is are three roots that exist, one always at the origin. Above the transcritical boundary there are 3 equilibrium points, a negative one, an unstable one at the origin, and a positive stable one. In the rest of the yellow region there is a stable equilibrium point at zero, and an unstable positive equilibrium point.

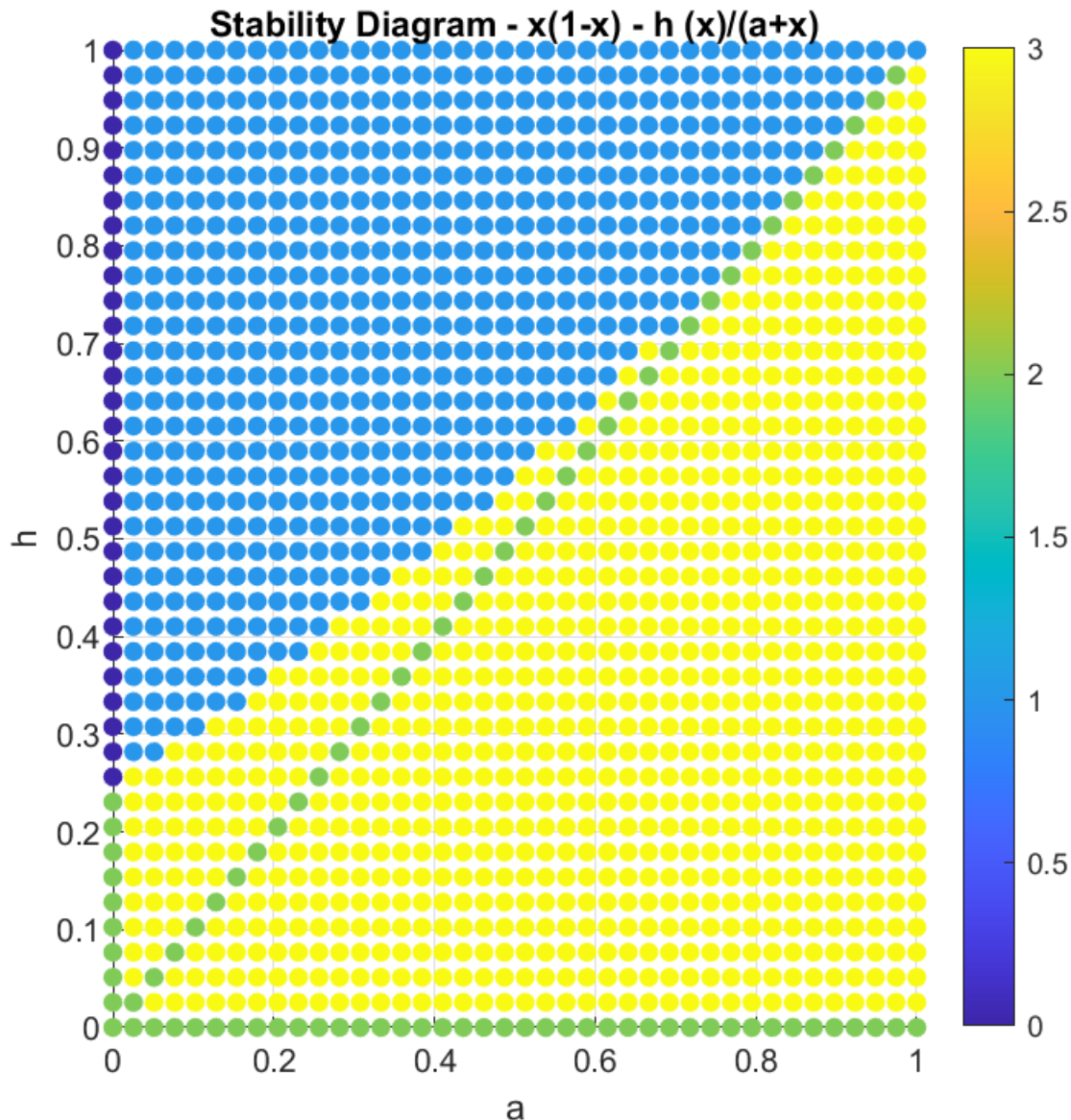


Figure 3: Stability diagram showing the roots of the system for various parameters a and h .

4 Problem 4 - K 3.8

Problem: Let the following system be defined:

$$\begin{aligned}\dot{x}_1 &= -x_1 + \frac{2x_2}{1+x_2^2}, \quad x_1(0) = a \\ \dot{x}_2 &= -x_2 + \frac{2x_1}{1+x_1^2}, \quad x_2(0) = b\end{aligned}\tag{32}$$

Show that this system has a unique solution for all $t \geq 0$.

Solution:

The system is known to be continuous on its domain. It is also apparent that both functions are differentiable, which results in a Jacobian of

$$\begin{bmatrix} -1 & \frac{2}{x_2^2+1} - \frac{4x_2^2}{(x_2^2+1)^2} \\ \frac{2}{x_1^2+1} - \frac{4x_1^2}{(x_1^2+1)^2} & -1 \end{bmatrix}\tag{33}$$

which indicates the systems dynamics are both differentiable and differential bounded. This also implies that the system is globally Lipschitz continuous, therefore a unique solution exists for $t \geq 0$.

5 Problem 5 - K 3.13

Problem: Let the following system be defined:

$$\begin{aligned}\dot{x}_1 &= \tan^{-1}(ax_1) - x_1x_2 \\ \dot{x}_2 &= bx_1^2 - cx_2\end{aligned}\tag{34}$$

Derive the sensitivity equations for the parameters vary from their nominal values of $a_0 = 1$, $b_0 = 0$, and $c_0 = 1$. Then simulate the sensitivity equations and the time dependence for the initial conditions of $x_1(0) = 1$ and $x_2(0) = -1$.

Solution:

5.1 Part a - Sensitivity Calculation

Let the following be defined:

$$\mu = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

Let the trajectory $x(\mu, t)$ be defined with regards to parameter changes as:

$$x(\mu, t) = x(\bar{\mu}, t) + \left. \frac{\partial x}{\partial \mu} \right|_{\bar{\mu}} \tilde{\mu}\tag{35}$$

where $\tilde{\mu} = \mu - \bar{\mu}$.

It can also be defined by its nonlinear definition as:

$$x(\mu, t) = x_0 + \int_0^t \dot{x}(x(\mu, \tau), \mu, \tau) d\tau\tag{36}$$

The sensativity to the parameters can then be formulated

$$S(t) = \frac{\partial x}{\partial \mu} = 0 + \frac{\partial}{\partial \mu} \int_0^t f(x(\mu, \tau), \mu, \tau) d\tau\tag{37}$$

$$= \int_0^t \frac{\partial}{\partial \mu} f(x(\mu, \tau), \mu, \tau) d\tau\tag{38}$$

$$= \int_0^t \frac{\partial f}{\partial x} \frac{\partial x}{\partial \mu} \frac{\partial f}{\partial \mu} d\tau\tag{39}$$

which can be clculated as jacobians of f resulting in

$$S(t) = \int_0^t A(\tau)S(\tau) + B(\tau) d\tau\tag{40}$$

where the matrices $A(\tau)$ and $B(\tau)$ are the Jacobians with respect to x and μ respectively:

$$\begin{aligned}A(\tau) &= \left. \frac{\partial f}{\partial x} \right|_{\bar{\mu}} & B(\tau) &= \left. \frac{\partial f}{\partial \mu} \right|_{\bar{\mu}} \\ &= \begin{bmatrix} -x_2 + \frac{x_1}{x_1^2 + 1} & -x_1 \\ 0 & 1 \end{bmatrix} & &= \begin{bmatrix} \frac{x_1}{x_1^2 + 1} & 0 & 0 \\ 0 & x_1^2 & -x_2 \end{bmatrix}\end{aligned}\tag{41}$$

Finally, the evolution of sensitivity over time can be found using the Leibnitz Formula, resulting in

$$\frac{ds(t)}{dt} = A(t)S(t) + B(t)\tag{42}$$

5.2 Part b - Simulation

The MATLAB code in AppendixA simulates and plots the states and sensitivities for each of the parameters. These plots can be seen in Figure4.

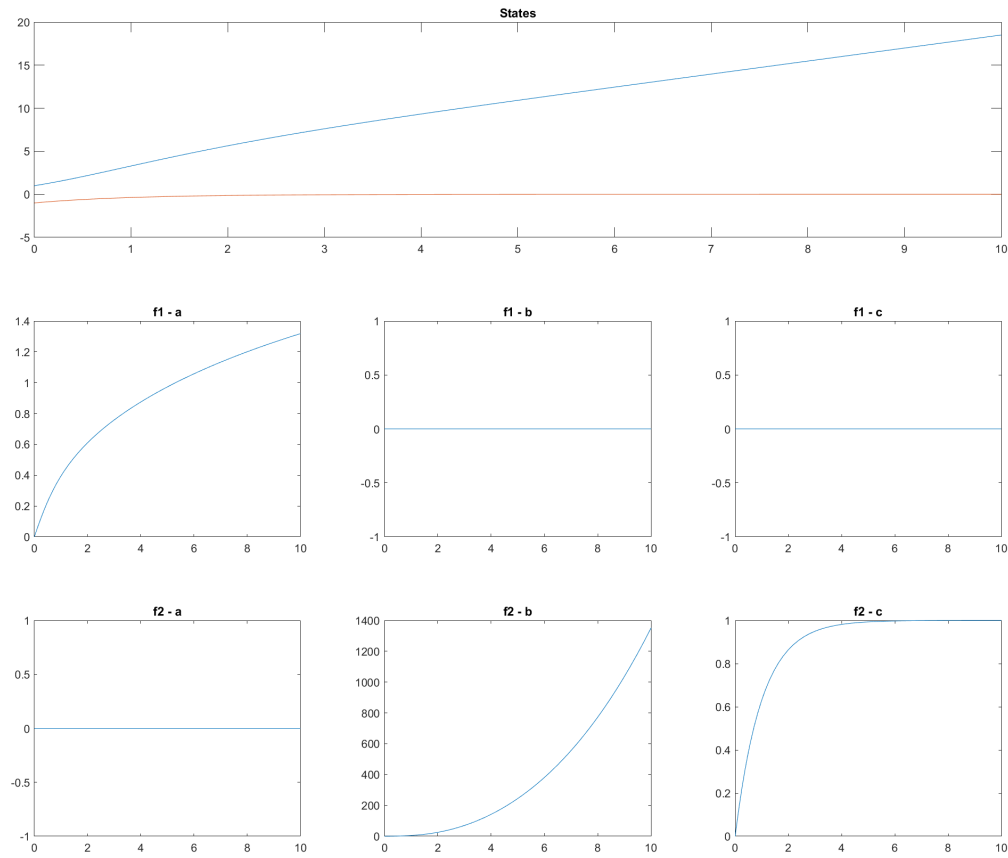


Figure 4: Simulation for Problem 5 with the evolution of sensitivities to parameters.

A MATLAB Code:

All code I write in this course can be found on my GitHub repository:

<https://github.com/jonaswagner2826/MECH6313>

Script 1: MECH6313_HW3

```
1 %% MECH6313 - HW 3
2 clear
3 close all
4
5 pblm1 = true;
6 pblm2 = false;
7 pblm3 = false;
8 pblm4 = false;
9 pblm5 = false;
10
11
12 if pblm1
13 %% Problem 1
14 syms x_1 x_2
15 x_1_dot = -x_1 + x_2;
16 x_2_dot = (x_1^2)/(1 + x_1^2) - 0.5 * x_2;
17 x_dot = [x_1_dot; x_2_dot]
18
19 % part a
20 syms x_1_bar x_2_bar
21
22 x_1_bar_dot = subs(x_1_dot, [x_1, x_2], [x_1_bar + 1, x_2_bar + 1])
23 x_2_bar_dot = subs(x_2_dot, [x_1, x_2], [x_1_bar + 1, x_2_bar + 1])
24
25 x_bar = [x_1_bar; x_2_bar]
26 x_bar_dot = [x_1_bar_dot; x_2_bar_dot]
27
28 % part b
29 % Linearize
30 A_sym = jacobian(x_bar_dot, x_bar)
31 A = subs(A_sym, [x_1_bar, x_2_bar], [0,0])
32
33 [T1, eig_A] = eig(A)
34
35 % Transform
36 syms y_sym z_sym
37 assume(y_sym,'real')
38 assume(z_sym,'real')
```

```

39 x_bar_sub = T1 * [y_sym; z_sym];
40 y_dot = subs(x_1_bar_dot, [x_1_bar, x_2_bar], x_bar_sub');
41 z_dot = subs(x_2_bar_dot, [x_1_bar, x_2_bar], x_bar_sub');
42
43 % G function Definitions
44 g1 = y_dot;
45 g2 = z_dot + 3/2 * z_sym;
46 'g1'
47 pretty(g1)
48 'g2'
49 pretty(g2)
50
51 % Coefficients of eigenvalue matrix
52 A1 = eig_A(1,1);
53 A2 = eig_A(2,2);
54
55 % w_dot substitution (from definition equation)
56 syms h_sym dh_sym
57 w_dot = A2 * h_sym + subs(g2,z_sym,h_sym) == dh_sym * (A1 * y_sym + subs(g1,z_sym,h_sym))
    ;
58
59 % Taylor's Series approximation of manifold
60 syms h2
61 h = h2 * y_sym^2;
62 dh = diff(h,y_sym);
63
64 % Attempting to solve for the h2 value...
65 w_dot = (subs(w_dot, [h_sym, dh_sym], [h, dh]));
66 'w_dot'
67 pretty(w_dot)
68 'How do I solve this???'
69 % solve(w_dot,h2)
70
71 % w_dot_soln = expand(w_dot);
72 % w_dot_soln = subs(w_dot_soln, y_sym^4, 0);
73 % w_dot_soln = subs(w_dot_soln, y_sym^3, 0);
74 % syms y2
75 % w_dot_soln = subs(w_dot_soln, y_sym^2, y2);
76 % w_dot_soln = subs(w_dot_soln, y_sym, 0);
77 % 'w_dot_soln'
78 % pretty(w_dot_soln)
79 %
80 % syms h2y2

```

```

81 % w_dot_soln = subs(w_dot_soln, h2*y2, h2y2)
82 % solve(w_dot_soln == 0,h2y2)
83
84 % Part 1d
85 'Part 1d'
86 'x_dot'
87 pretty(x_dot)
88 % f definition
89 f = matlabFunction(x_dot);
90
91 % Fig def
92 fig = figure('position',[0,0,1500,1200]);
93
94 % Quiver Plot
95 [X,Y] = meshgrid([-20:2:20]);
96 temp = f(X,Y);
97 U = temp(1:(size(X,1)),:);
98 V = temp((size(X,1)+1):2*size(X,1),:);
99 quiver(X,Y,U,V)
100 hold on
101
102 % Phase Plots
103 [X,Y] = meshgrid([-20:5:20]);
104 X_0 = [X(:),Y(:)];
105 T = [0,10];
106 for i = 1:size(X_0,1)
107     x_0 = X_0(i,:);
108     [~,y] = ode45(@(t,y) f(y(1),y(2)),T,x_0);
109     plot(y(:,1),y(:,2),'r')
110 end
111
112
113 % Z and Y Axes
114 T_inv = inv(T1);
115 syms x1_sym x2_sym
116 y_axis = matlabFunction(solve(0 == T_inv(1,:) * [x1_sym; x2_sym],x2_sym));
117 z_axis = matlabFunction(solve(0 == T_inv(2,:) * [x1_sym; x2_sym],x2_sym));
118
119 X1 = [-15:0.2:15];
120 X2_y = y_axis(X1) + 1; %adjust from xbar to x
121 X2_z = z_axis(X1) + 1; %adjust from xbar to x
122 X1 = X1 + 1; %adjust from xbar to x
123

```

```

124
125 plot(X1,X2_y, 'k', 'LineWidth',3)
126 plot(X1,X2_z, 'k', 'LineWidth',3)
127
128 hold off
129
130 title('Phase Portrait for Problem 1')
131 xlabel('x1')
132 ylabel('x2')
133
134 % Save figure
135 saveas(fig,fullfile([pwd '\\ 'HW3' '\\ 'fig'],'pblm1.png'))
136
137
138
139
140 end
141
142
143 if pblm2
144 %% Problem 2
145 syms x h
146 dx(x,h) = x * (1-x) - h;
147 f = matlabFunction(dx);
148
149 fig = figure('position',[0,0,1000,1000]);
150
151 [X,Y] = meshgrid([0:0.1:1.5],[0:0.05:0.5]);
152 U = f(X,Y);
153 V = 0 * U;
154 q = quiver(X,Y,U,V);
155 q.AutoScaleFactor = 2;
156
157 title('Vector Field Plot: dx = x(1-x) - h')
158 xlabel('x')
159 ylabel('h')
160
161 saveas(fig,fullfile([pwd '\\ 'HW3' '\\ 'fig'],'pblm2.png'))
162
163 end
164
165
166 if pblm3

```

```

167 %% Problem 3
168 syms x h a
169 dx(x,h,a) = x * (1-x) - h * ((x)/(a+x));
170 f = matlabFunction(dx);
171
172 x1 = linspace(0,1,40);
173 y1 = linspace(0,1,40);
174 [X,Y] = meshgrid(x1,y1);
175 %method of finding num roots:
176 realRoots = solve(0==dx(x,0,0),x,'Real',true);
177 numRoots = size(realRoots,1);
178
179 for i = 1:size(X,1)
180     for j = 1:size(X,2)
181         a = X(i,j);
182         h = Y(i,j);
183         Z(i,j) = size(solve(0==f(x,h,a),x,'Real',true),1);
184     end
185 end
186
187 %Ploting
188 fig = figure('position',[0,0,700,700]);
189 scatter3(X(:),Y(:),Z(:),[],Z(:),'filled')
190 view(2)
191 colorbar
192
193 title('Stability Diagram -  $x(1-x) - h(x)/(a+x)$ ')
194 xlabel('a')
195 ylabel('h')
196
197 % saveas(fig,fullfile([pwd '\\ 'HW3' '\\ 'fig'],'pblm3.png'))
198
199 end
200
201
202 if pblm4
203 %% Problem 4
204 syms x1 x2
205 x1_dot = -x1 + (2*x2)/(1 + x2^2);
206 x2_dot = -x2 + (2*x1)/(1 + x1^2);
207 f = [x1_dot; x2_dot];
208 'f'
209 pretty(f)

```



```

210 df = jacobian(f);
211 'jacobian'
212 pretty(df)
213 end
214
215
216 if pblm5
217 %% Problem 5
218 syms x1 x2 a b c
219 x1_dot = atan(a * x1) - x1 * x2;
220 x2_dot = b * x1^2 - c * x2;
221 x_dot = [x1_dot; x2_dot];
222 'x_dot'
223 pretty(x_dot)
224
225 x = [x1; x2];
226 'x'
227 pretty(x)
228
229 mu = [a; b; c];
230 mu_bar = [1; 0; 1];
231 'mu'
232 pretty(mu)
233
234
235 A_tau = jacobian(x_dot, x)
236 B_tau = jacobian(x_dot, mu)
237
238
239 sys_func = @pblm5_func;
240
241 T = [0,10];
242 x_0 = [1,-1, 0,0,0,0,0,0]';
243
244 [t,y] = ode45(@(t,y) sys_func(t,y,mu_bar,A_tau,B_tau),T,x_0);
245
246 y_states = y(:,[1,2]);
247 y_a = y(:,[3,6]);
248 y_b = y(:,[4,7]);
249 y_c = y(:,[5,8]);
250
251 fig = figure('position',[0,0,1500,1200]);
252 subplot(3,3,[1:3])

```

```

253 plot(t,y_states)
254 title('States')
255
256 titles = ["f1 - a", "f1 - b", "f1 - c", "f2 - a", "f2 - b", "f2 - c"];
257 for i = 3:8
258     subplot(3,3,i+1)
259     plot(t,y(:,i))
260     title(titles(i-2))
261 end
262
263 saveas(fig,fullfile([pwd '\\' 'HW3' '\\' 'fig'],'pblm5.png'))
264
265 end
266
267
268
269
270
271 %% Local Functions
272 function dx = pblm5_func(t, x, parms, A, B)
273     % pblm5 function
274     arguments
275         t (1,1) = 0;
276         x (8,1) = zeros(8,1); %state and 6 sensitivities
277         parms = false;
278         A = 0;
279         B = 0;
280     end
281
282     if parms == false
283         a = 1;
284         b = 0;
285         c = -1;
286     else
287         a = parms(1);
288         b = parms(2);
289         c = parms(3);
290     end
291
292     % Variable Decode
293     x1 = x(1);
294     x2 = x(2);
295     S = zeros(2,3);%[x(3), x(4), x(5); x(6), x(7), x(8)];

```

```

296
297 % State Upadate Eqs
298 x1_dot = atan(a * x1) - x1 * x2;
299 x2_dot = b * x1^2 - c * x2;
300 S_dot = subs(A * S + B);
301
302 % Variable Encode
303 dx = x;
304 dx(1) = x1_dot;
305 dx(2) = x2_dot;
306 dx(3) = S_dot(1,1);
307 dx(4) = S_dot(1,2);
308 dx(5) = S_dot(1,3);
309 dx(6) = S_dot(2,1);
310 dx(7) = S_dot(2,2);
311 dx(8) = S_dot(2,3);
312 end

```