## Code/Plots

The problems were integrated using solve\_ivp from the scipy library in Python.

#### 12.2.2.1)

```
#!/bin/python
  ##-----
  # Libraries
  import matplotlib.pyplot as plt
  import numpy as np
  from scipy.integrate import solve_ivp
  from plot import plot
  ##-----
  # Functions
13
14
  ##-----
15
16
  def model(t,x):
17
    input
19
      t: time step
20
      x: state
21
    output
       dx: differential output of ode
    # Extract states
26
    x1, x2, u = x
27
28
    # Calculate z
29
    z1 = x1
    z2 = x1 + x2
    z = np.array([[z1], [z2]])
    # Calculate control
    k = np.array([10, 10])
    u = -3*(x1**3)*x2 + 2*x1 + x2 - k0z
    # Calculate ODEs
    xd1 = x1 + x2
    xd2 = 3*(x1**2)*x2 + x1 + u
40
41
    # Bundle ODEs
42
    dx = [xd1, xd2, u]
43
    return dx
  ##-----
  # Script
  # Plotting/Calculating variables
```

```
## Time horizon
t0 = 0.0
tf = 15.0
t = [t0, tf]

## Initial conditions
x0 = [1, 0.1, 0]

## Solve ode model
sol = solve_ivp(model, t, x0, method='RK45', dense_output=True)

## Plot solution
plot(sol, tf=tf, title="12.2.2.1", legend=['x1', 'x2', 'u'])
```

#### 12.2.2.2)

```
#!/bin/python
3 ##-----
4 # Libraries
  import matplotlib.pyplot as plt
  import numpy as np
  from scipy.integrate import solve_ivp
10 from plot import plot
11
12 ##-----
13 # Functions
14
15 ##-----
16 #
def model(t,x):
18
    input
19
     t: time step
20
      x: state
21
22
    output
      dx: differential output of ode
23
24
    0.00
    # Extract state
    x1, x2, x3, u = x
            = np.array([x1,x2,x3])
    # Calculate control
    k = np.array([10, 10, 5])
    u = -k0x[0:4]
    # Calculate ODEs
34
    xd1 = x1 + x2
35
    xd2 = x1*x2**2 - x1 + x3
36
    xd3 = u
37
    # Bundle ODEs
    dx = [xd1, xd2, xd3, u]
41
    return dx
42
  ##-----
  # Script
  # Plotting/Calculating variables
49 ## Time horizon
50 t0 = 0.0
51 tf = 15.0
52 t = [t0, tf]
54 ## Initial conditions
```

```
55  x0 = [0.5, 0.1, 0.1, 0]
56
57
58  # Solve ode model
59  sol = solve_ivp(model, t, x0, method='RK45', dense_output=True)
60
61  # Plot solution
62  plot(sol, tf=tf, title="12.2.2.2", legend=['x1', 'x2', 'u'])
```

#### 14.31)

```
#!/bin/python
3 ##-----
4 # Libraries
  import matplotlib.pyplot as plt
  import numpy as np
  from scipy.integrate import solve_ivp
10 from plot import plot
11
12 ##-----
13 # Functions
14
15 ##-----
16 #
def model(t,x):
18
    input
19
    t: time step
20
      x: state
21
22
    output
      dx: differential output of ode
23
24
    0.00
    # Extract states
    x1, x2, u = x
    a = 0
    # Calculate z
    z = x1 + x2 + a + (x1 - a**1/3)**3 + x1
    # Calculate control
    u = -x1 - x2 - 3*(x1**2)*x2 - z
34
35
    # Calculate ODEs
36
37
    xd1 = x2 + a + (x1 - a**1/3)**3
    xd2 = x1 + u
    # Bundle ODEs
    dx = [xd1, xd2, u]
42
43
    return dx
44
  ##-----
  # Script
47
48
49 # Plotting/Calculating variables
50
51 ## Time horizon
52 t0 = 0.0
53 tf = 8.0
t = [t0, tf]
```

```
## Initial conditions

x0 = [0.5, 0.6, 0]

8

9

60 # Solve ode model

sol = solve_ivp(model, t, x0, method='RK45', dense_output=True)

22

63 # Plot solution
64 plot(sol, tf=tf, title="14.31", legend=['x1', 'x2', 'u'])
```

### 14.34)

```
1 ##-----
2 # Libraries
  import matplotlib.pyplot as plt
  import numpy as np
  from scipy.integrate import solve_ivp
  from plot import plot
  # Functions
11
14 #
def model(t,x):
     0.00
16
     input
17
      t: time step
18
       x: state
     output
20
        dx: differential output of ode
21
22
     0.00
23
     # Extract states
24
     x1, x2, x3, u = x
     # Calculate z
     z = x2
     # Calculate control
     u = -x1 + x2 + x1*x3 - z
     # Calculate ODEs
34
     xd1 = -x1 + x2
     xd2 = x1 - x2 - x1*x3 + u
35
     xd3 = x1 + x1*x2 - 2*x3
36
37
     # Bundle ODEs
     dx = [xd1, xd2, xd3, u]
     return dx
41
42
  ##-----
  # Script
  # Plotting/Calculating variables
  ## Time horizon
  t0 = 0.0
50 tf = 8.0
51 t = [t0, tf]
53 ## Initial conditions
x0 = [0.1, 0.2, 0.3, 0]
```

```
55

56

57  # Solve ode model

58  sol = solve_ivp(model, t, x0, method='RK45', dense_output=True)

59

60  # Plot solution

61  plot(sol, tf=tf, title="14.34", legend=['x1', 'x2', 'x3', 'u'])
```

#### 14.15 a)

```
#!/bin/python
3 ##-----
  # Libraries
  import matplotlib.pyplot as plt
  import numpy as np
  from scipy.integrate import solve_ivp
10 from plot import plot
11
12 ##=----
13 # Functions
14
15 ##-----
16 #
def model(t,x):
18
    input
19
    t: time step
20
21
      x: state
22
    output
      dx: differential output of ode
23
24
    # Extract states
    x1, x2, u = x
    # Calculate control
    k = 0.25
    u = -k*x2
    # Calculate ODEs
33
34
    xd1 = x2
35
    xd2 = -x1**3 + u
36
37
    # Bundle ODEs
    dx = [xd1, xd2, u]
    return dx
41
42
  ##-----
  # Script
  # Plotting/Calculating variables
  ## Time horizon
48
  t0 = 0.0
50 tf = 100
t = [t0, tf]
53 ## Initial conditions
x0 = [0.5, 0.6, 0]
```

```
55
56
57 # Solve ode model
58 sol = solve_ivp(model, t, x0, method='RK45', dense_output=True)
59
60 # Plot solution
61 plot(sol, tf=tf, title="14.15_a", legend=['x1', 'x2', 'u'])
```

#### 14.15 b)

```
#!/bin/python
 # Libraries
  import matplotlib.pyplot as plt
  import numpy as np
  from scipy.integrate import solve_ivp
10 from plot import plot
11
12 ##=----
13 # Functions
14
15 ##-----
16 #
def model(t,x):
18
    input
19
    t: time step
20
      x: state
21
22
    output
      dx: differential output of ode
23
24
    # Extract states
    x1, x2, u = x
    # Calculate control
    k = 0.25
    u = -(2*k/np.pi)*np.arctan(x2)
    # Calculate ODEs
33
34
    xd1 = x2
35
    xd2 = -x1**3 + u
36
37
    # Bundle ODEs
    dx = [xd1, xd2, u]
    return dx
41
42
  ##-----
  # Script
  # Plotting/Calculating variables
  ## Time horizon
48
  t0 = 0.0
50 tf = 100
t = [t0, tf]
53 ## Initial conditions
x0 = [0.1, 0.1, 0.0]
```

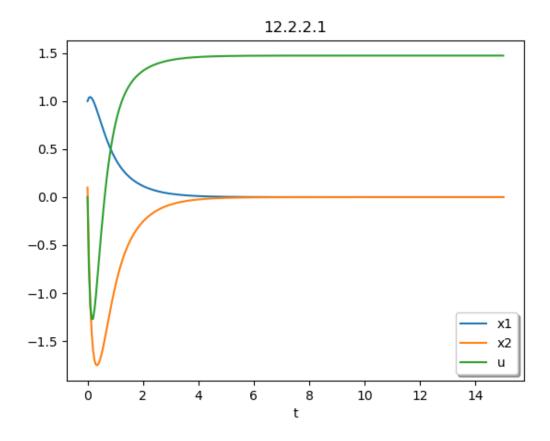
```
55
56
57 # Solve ode model
58 sol = solve_ivp(model, t, x0, method='RK45', dense_output=True)
59
60 # Plot solution
61 plot(sol, tf=tf, title="14.15_b", legend=['x1', 'x2', 'u'])
```

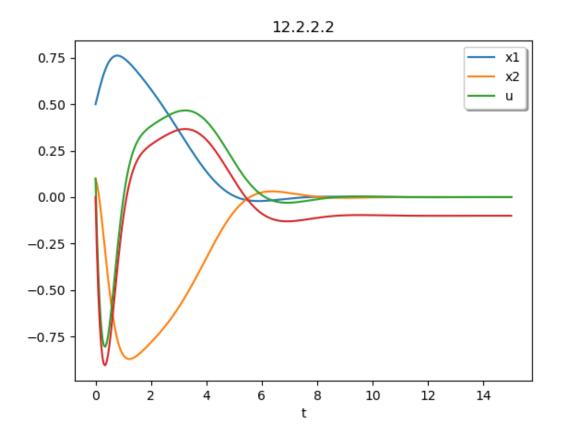
#### Plotter

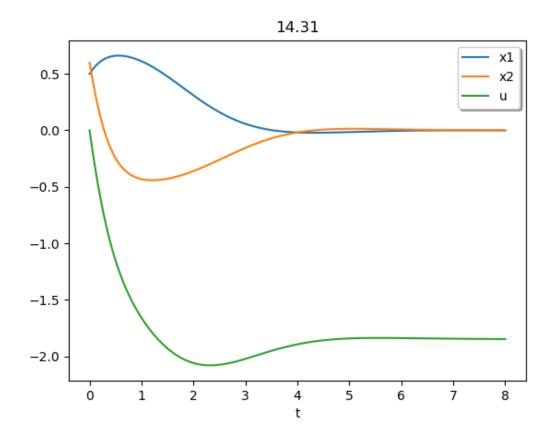
```
#!/bin/python
3 ##-----
4 # Libraries
  import matplotlib.pyplot as plt
   import numpy as np
9
10
   def plot(solution,
11
        tf : float = 15,
         x_{lbl} : str = "t",
         legend : np.array = ['x', 'y'],
14
         title : str = "Model",
15
         show_plot : bool = False):
16
      0.00
17
      Plotting module for ODE's.
18
      input
20
        solution: Solution from ode
21
22
      output
23
        NONE
24
      0.00
      # Extract
      t = np.linspace(0, tf, 300)
     z = solution.sol(t)
      # Plot
31
      plt.plot(t, z.T)
      plt.xlabel(x_lbl)
33
      plt.legend(legend, shadow=True)
34
      plt.title(title)
35
36
      # Determine whether to display or save plot
37
      if show_plot:
        plt.show()
      else:
40
         plt.savefig("img/"+title+".png")
41
42
      return
43
```

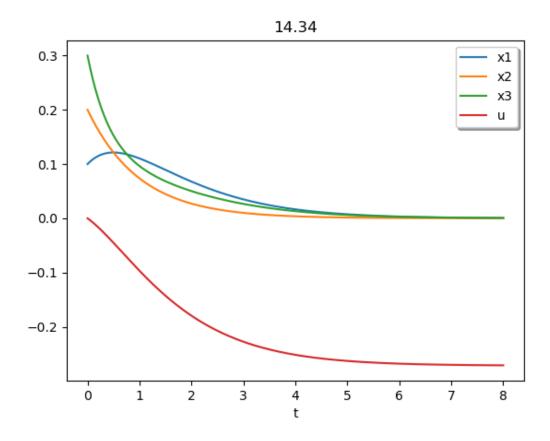
#### Script Runner

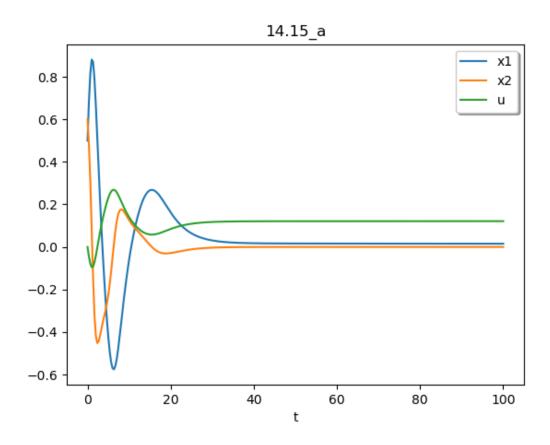
```
#!/bin/sh
   # Create a fancy little progress bar
   # https://stackoverflow.com/questions/238073/how-to-add-a-progress-bar-to-a-shell-script
   function ProgressBar
       # Process data
       let _progress=(${1}*100/${2}*100)/100
       let _done=(${_progress}*4)/10
       let _left=40-$_done
10
11
       # Build progressbar string lengths
       _fill=$(printf "%${_done}s")
       _empty=$(printf "%${_left}s")
14
15
       \ensuremath{\text{\#}}\xspace 1.2 Build progressbar strings and print the ProgressBar line
16
       # 1.2.1 Output example:
17
       # 1.2.1.1 Progress : [########################### 100%
       printf "\rProcessing $s: [${_fill// /#}${_empty// /-}] ${_progress}%% \r"
   }
20
21
   # Variables
   i=1
22
23
# Create list of scripts to be ran
   declare -a scripts=(`find . -type f -name "*.py" ! -name "plot.py"`)
# Loop through each script
28 for s in "${scripts[0]}"
29
       ## Update progress bar
30
       ProgressBar $i ${#scripts[@]}
31
       ## Process script
33
       python $s
34
35
       ## Update Index
36
       i=$((i+1))
37
   done
40 printf '\nFinished!\n'
```

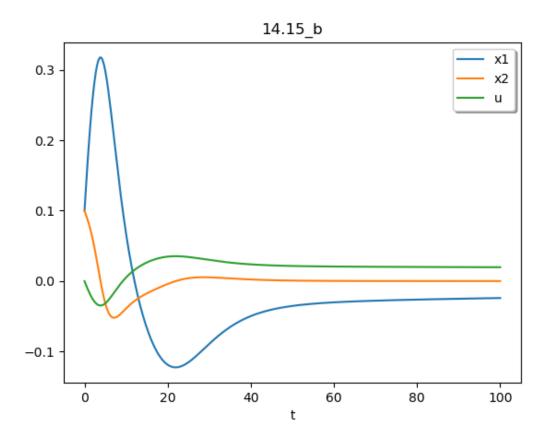












# A note on 14-15-a and 14-15-b

Comparing the two plots, it can be noted that a has a better state response with more control applied. b converges slower, but has less overshoot with and less control applied.