```
2 # IMPORTS
4 import numpy as np
5 import os
6 import matplotlib.pyplot as plt
7 import matplotlib as mpl
8 from kalman import kalman filter
10 # Variable declaration
12 isPlotReqd = True
13 isPlotPdf = True
15 # Settings for plot
17 if (True == isPlotRegd):
18
   if (True == isPlotPdf):
19
    mpl.use('pdf')
20
    fig width = 3.487
    fig height = fig_width / 1.618
21
22
    rcParams = {
23
     'font.family' 'serif',
24
     'font.serif': 'Times New Roman',
25
     'text.usetex': True,
26
     'xtick.labelsize': 8,
27
     'ytick.labelsize': 8,
28
     'axes.labelsize': 8,
29
     'legend.fontsize': 8,
30
     'figure.figsize': [fig width, fig height]
31
32
    plt.rcParams.update(rcParams)
34 # CODE STARTS HERE
37 # Parameters
39 x 0 = \text{np.array}([[650.], [250.]])
40 x 0 hat = np.array([[600.], [200.]])
41 P_0_{hat} = np.array([[500., 0.], [0., 200.]])
42 Q = \text{np.array}([[0., 0.], [0., 10.]])
43 H = np.array([[1, 0]])
44 R = 10.
45 F = np.array([[0.5, 2], [0, 1]])
46 G = np.zeros(x \ 0 \ hat.shape)
48 # Time steps
50 num step = 10
51 t = np.linspace(0, num step, num=num step + 1)
53 # Dynamic system simulation
55 class dynamic_system():
   def init (self, num step, x 0, F, G, H, Q, R):
56
57
    self.num step = num step
58
    self.t = np.linspace(0, num step, num=num step + 1)
59
    self.x 0 = x = 0
60
    self.F = F
    self.G = G
61
```

```
62
       self.H = H
63
       self.Q = Q
       self.R = R \# Right not supporting scalar R
64
65
     def get simulated measurements(self):
66
       w f = np.zeros(self.t.shape)
67
       w f[1:] = \text{np.sqrt}(\text{self.Q}[1, 1]) * \text{np.random.normal}(0, 1, (\text{self.t.shape}[0] - 1,))
68
       w k = np.vstack((np.zeros(self.t.shape), w f))
69
70
       x true = np.zeros((self.x 0.shape[0], self.t.shape[0]))
71
       x \text{ true}[:, 0] = \text{self.} x \ 0[:, 0]
72
       y_k = np.zeros((1, self.t.shape[0]))
       for step in range(self.num step):
73
74
         x \text{ true}[:, \text{step} + 1] = \text{self.F} @ x_\text{true}[:, \text{step}] + w_k[:, \text{step}]
75
         y k[0, step + 1] = self.H @x true[:, step + 1] + np.sqrt(self.R) * np.random.normal()
76
       return v k
78 # Kalman filter prediction
80 ds = dynamic system(num step=num step, x 0=x 0, F=F, G=G, H=H, Q=Q, R=R)
81 y k = ds.get simulated measurements()
82 kf = kalman filter(init state=x 0 hat, init est err=P 0 hat, Q=Q, R=R, F=F, G=G, H=H, y k=y k,
   num step=num step)
83 kf.run kalman()
84 kf.run fb smoother(num meas=10, est idx=-1)
85 P f = kf.get est error cov()
86 P fb = kf.get est error cov fb sm()
88 # Plot of cov of estimation error: population
90 if (True == isPlotReqd):
91
     92
     # Configure axis and grid
93
     94
     fig = plt.figure()
95
     ax = fig.add subplot(111)
96
     fig.subplots adjust(left=.15, bottom=.16, right=.99, top=.97)
97
98
     ax.set axisbelow(True)
99
     ax.minorticks on()
100
     ax.grid(which='major', linestyle='-', linewidth='0.5')
     ax.grid(which='minor', linestyle=''-.", linewidth='0.5')
101
102
103
     ax.plot(t, P f[0, 0, :], color='r', label='Population error cov: forward')
104
     ax.plot(t, P fb[0, 0, :], color='g', label='Population error cov: smoothed')
105
106
     ax.set xlabel(r'time', fontsize=8)
107
     ax.set ylabel(r'standard deviation', fontsize=8)
108
109
     plt.legend()
110
     if (True == isPlotPdf):
111
       if not os.path.exists('./generatedPlots'):
112
         os.makedirs('generatedPlots')
113
       fig.savefig('./generatedPlots/q5 population cov.pdf')
114
115
       plt.show()
117 # Plot of cov of estimation error: food
119 if (True == isPlotRegd):
120
     121
     # Configure axis and grid
```

```
122
      123
      fig = plt.figure()
124
      ax = fig.add subplot(111)
125
      fig.subplots adjust(left=.15, bottom=.16, right=.99, top=.97)
126
127
      ax.set axisbelow(True)
128
      ax.minorticks on()
129
      ax.grid(which='major', linestyle='-', linewidth='0.5')
130
      ax.grid(which='minor', linestyle="-.", linewidth='0.5')
131
132
      ax.plot(t, P f[1, 1, :], color='b', label='food error cov: forward')
133
      ax.plot(t, P fb[1, 1, :], color='m', label='food error cov: smoothed')
134
135
      ax.set xlabel(r'time', fontsize=8)
136
      ax.set ylabel(r'standard deviation', fontsize=8)
137
138
      plt.legend()
139
      if (True == isPlotPdf):
140
        if not os.path.exists('./generatedPlots'):
141
          os.makedirs('generatedPlots')
        fig.savefig('./generatedPlots/q5_food cov.pdf')
142
143
      else:
144
        plt.show()
146 # (b)Percentage change of cov estimates at initial time due to smoothing
148 improvement popu = ((P_f[0, 0, 0] - P_fb[0, 0, 0])/P_fb[0, 0, 0])*100.
149 improvement_food = ((P_f[1, 1, 0] - P_fb[1, 1, 0])/P_fb[1, 1, 0])*100.
150 print('Percentage change of initial estimation error variance due to smoothing: population: ',
    improvement popu)
151 print('Percentage change of initial estimation error variance due to smoothing: food: ',
    improvement food)
153 # (c) Numerical estimate of error cov at initial time
155 n run = 100
156 init cov est popu = np.zeros((n run,))
157 init cov est food = np.zeros((n run,))
158 for run idx in range(n run):
      y k = ds.get simulated measurements()
159
160
      kf.y k = y k
161
      kf.run fb smoother(num meas=10, est idx=-1)
162
      P fb est = kf.get est error cov fb sm()
163
      init\_cov\_est\_popu[run\_idx] = P\_fb\_est[0, 0, 0]
      init cov est food[run idx] = P fb est[1, 1, 0]
164
165 print('Numerical estimate of initial population error cov = ', np.mean(init_cov_est_popu))
166 print('Theoretical estimate of initial population error cov = ', P fb[0, 0, 0])
    print('Numerical estimate of initial food supply error cov = ', np.mean(init cov est food))
168 print('Theoretical estimate of initial food supply error cov = ', P fb[1, 1, 0])
169
```