EE430 - Final Exam

Q1)

a) The ordinary differential equations of the dynamic model are given below.

$$\frac{d([X])}{dt} = k_2([XY])(t) + k_4([XE])(t) - k_1([X])(t)([Y])(t) - k_3([X])(t)([E])(t) - k_8([YE])(t)([X])(t)$$

$$\frac{d([Y])}{dt} = k_2([XY])(t) + k_6([YE])(t) - k_1([X])(t)([Y])(t) - k_5([Y])(t)([E])(t) - k_7([XE])(t)([Y])(t)$$

$$\frac{d([E])}{dt} = k_4([XE])(t) + k_6([YE])(t) + k_7([XE])(t)([Y])(t) + k_8([YE])(t)([X])(t) - k_3([X])(t)([E])(t)$$
$$-k_5([Y])(t)([E])(t)$$

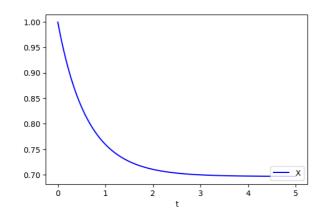
$$\frac{d([XY])}{dt} = k_1([X])(t)([Y])(t) + k_7([XE])(t)([Y])(t) + k_8([YE])(t)([X])(t) - k_2([XY])(t)$$

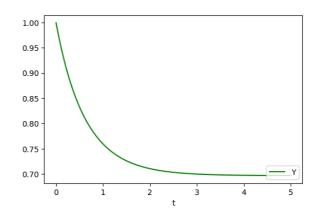
$$\frac{d([XE])}{dt} = k_3([X])(t)([E])(t) - k_4([XE])(t) - k_7([XE])(t)([Y])(t)$$

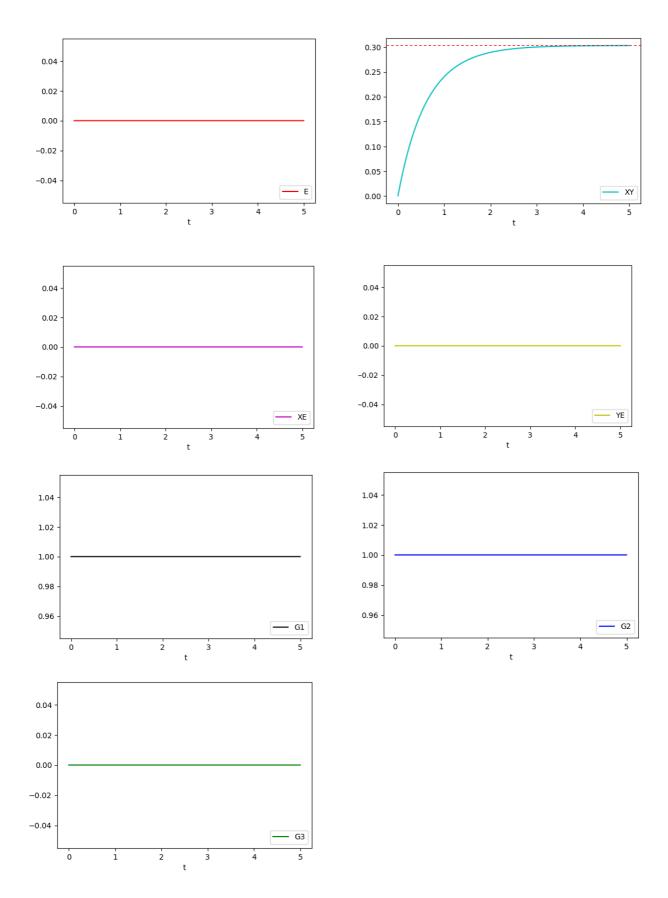
$$\frac{d([YE])}{dt} = k_5([Y])(t)([E])(t) - k_6([YE])(t) - k_8([YE])(t)([X])(t)$$

b) The transient behavior of all reactants and moiety groups are given below with initial conditions $[X]|_{t=0} = [Y]|_{t=0} = 1$ and $[E]|_{t=0} = [XE]|_{t=0} = [YE]|_{t=0} = [XY]|_{t=0} = 0$. In addition moeity groups are $G_1 = \{X, XE, XY\}, G_2 = \{Y, YE, XY\}, G_3 = \{E, XE, YE\}$. Besides, the k values have been chosen experimentally as $k_1 = 0.5$, $k_2 = 0.8$, $k_3 = 0.8$, $k_4 = 1.75$, $k_5 = 1.2$, $k_6 = 1.5$, $k_7 = 2.5$, $k_8 = 2.2$.

In order to ensure that the XY production in the case of E is greater than in the case of the absence of E, care was taken to ensure that k7 and k8 are greater than the value of k1.

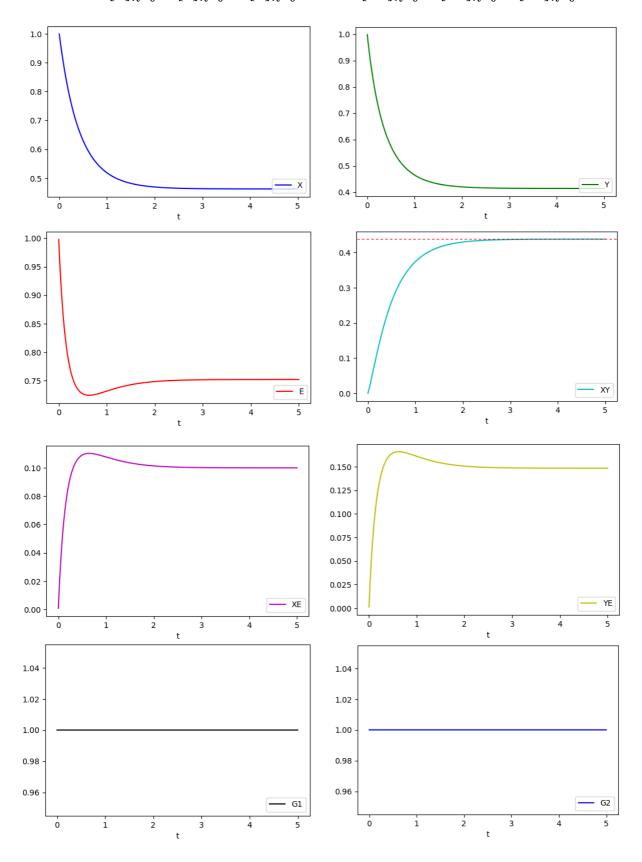


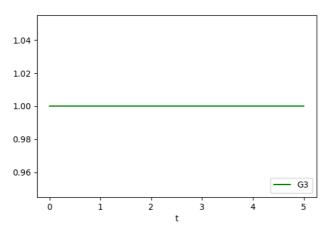




It has been observed that XE, YE are not produced due to the absence of E. The concentration of XY is about 0.30

c) The transient behavior of all reactants and moiety groups are given below with initial conditions $[X]|_{t=0} = [Y]|_{t=0} = [E]|_{t=0} = 1$ and $[XE]|_{t=0} = [YE]|_{t=0} = [XY]|_{t=0} = 0$.

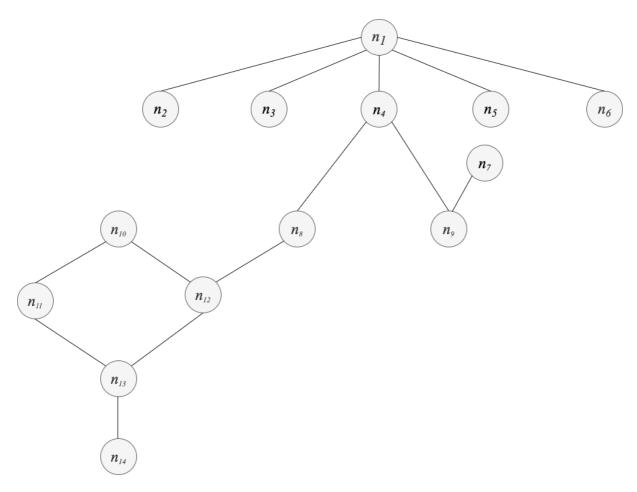




In the case of E, the XY concentration is about 0.44. In the presence of E, the concentration of XY is greater than in the absence of it.

Q2)

a) The graph of the undirected network is given below.



b) Degrees and cluster coefficients were calculated by counting the edges as in below. $\deg(n_1) = 5, \deg(n_2) = 1, \deg(n_3) = 1, \deg(n_4) = 3, \deg(n_5) = 1,$ $\deg(n_6) = 1, \deg(n_7) = 1, \deg(n_8) = 2, \deg(n_9) = 2,$ $\deg(n_{10}) = 2, \deg(n_{11}) = 2, \deg(n_{12}) = 3, \deg(n_{13}) = 3, \deg(n_{14}) = 1.$

$$C_1 = C_2 = C_3 = C_4 = C_5 = C_6 = C_7 = C_8 = C_9 = C_{10} = C_{11} = C_{12} = C_{13} = C_{14} = 0$$

The value of e is 0 for all nodes since there is no edge between the neighbors of the nodes. If the value of e in the cluster coefficient formula given below is 0, the cluster coefficients for all nodes are 0.

$$C = 2 \frac{e}{\deg(n_i)(\deg(n_i) - 1)}$$

c) The adjacency matrix of the system id given below.

To get the eigenvector centrality of the graph, computer algorithms were used.(Appendix 2). For the maximum eigenvalue, the eigenvectors are given in below.

$$v = \begin{bmatrix} 0.5317 \\ 0.218 \\ 0.218 \\ 0.424 \\ 0.218 \\ 0.218 \\ 0.086 \\ 0.294 \\ 0.209 \\ 0.190 \\ 0.172 \\ 0.292 \\ 0.229 \\ 0.094 \end{bmatrix}$$

Here are the nodes in a decreasing order of centrality;

$$n_4, n_{12}, n_2, n_3, n_5, n_6 n_{11}, n_{14}, n_7, n_{10}, n_9, n_{13}, n_8, n_1.$$

Appendix I

```
import numpy as np
import math
import matplotlib.pyplot as plt
dt = 0.001
t = np.arange(0, 5, dt)
X = np.ndarray((len(t)))
Y = np.ndarray((len(t)))
E = np.ndarray((len(t)))
XY = np.ndarray((len(t)))
XE = np.ndarray((len(t)))
YE = np.ndarray((len(t)))
G1 = np.ndarray((len(t)))
G2 = np.ndarray((len(t)))
G3 = np.ndarray((len(t)))
k1 = 0.5
k2 = 0.8
k3 = 0.8
k4 = 1.75
k5 = 1.2
k6 = 1.5
k7 = 2.5
k8 = 2.2
X i = 1.0
Y i = 1.0
E i = 0.0
XY i = 0.0
XE i = 0.0
YE i = 0.0
cind = 0
for ct in t:
  dXdt = k2*XY i + k4*XE i - k1*X i*Y i - k3*X i*E i - k8*YE i*X i
  dYdt = k2*XY i + k6*YE i - k1*X i*Y i - k5*Y i*E i - k7*XE i*Y i
  dEdt = k4*XE i + k6*YE i + k7*XE i*Y i + k8*YE i*X i - k3*X i*E i -
k5*Y i*E i
  dXYdt = k1*X i*Y i + k7*XE i*Y i + k8*YE i*X i - k2*XY i
  dXEdt = k3*X_i*E_i - k4*XE_i - k7*XE_i*Y_i
  dYEdt = k5*Y i*E i - k6*YE i - k8*YE i*X i
 X n = X_i + dt * dXdt
 Y_n = Y_i + dt * dYdt
 E n = E i + dt * dEdt
```

```
XY_n = XY_i + dt * dXYdt
  XE n = XE i + dt * dXEdt
  YE n = YE i + dt * dYEdt
  X[cind] = X n
  Y[cind] = Y n
  E[cind] = E n
  XY[cind] = XY_n
  XE[cind] = XE n
  YE[cind] = YE n
  G1[cind] = format(X n + XY n + XE n, '.2f')
  G2[cind] = format(Y_n + XY_n + YE_n, '.2f')
  G3[cind] = format(E_n + XE_n + YE_n, '.2f')
 X i = X_n
 Y_i = Y_n
 E i = E n
 XY_i = XY_n
 XE i = XE n
 YE_i = YE_n
  cind +=1
plt.figure(dpi=100)
plt.plot(t, X, 'b', label='X')
plt.xlabel('t')
plt.legend(loc='lower right')
plt.show()
plt.figure(dpi=100)
plt.plot(t, Y, 'g', label='Y')
plt.xlabel('t')
plt.legend(loc='lower right')
plt.show()
plt.figure(dpi=100)
plt.plot(t, E, 'r', label='E')
plt.xlabel('t')
plt.legend(loc='lower right')
plt.show()
plt.figure(dpi=100)
plt.plot(t, XY, 'c', label='XY')
plt.xlabel('t')
plt.legend(loc='lower right')
```

```
plt.show()
plt.figure(dpi=100)
plt.plot(t, XE, 'm', label='XE')
plt.xlabel('t')
plt.legend(loc='lower right')
plt.show()
plt.figure(dpi=100)
plt.plot(t, YE, 'y', label='YE')
plt.xlabel('t')
plt.legend(loc='lower right')
plt.show()
plt.figure(dpi=100)
plt.plot(t, G1, 'k', label='G1')
plt.xlabel('t')
plt.legend(loc='lower right')
plt.show()
plt.figure(dpi=100)
plt.plot(t, G2, 'b', label='G2')
plt.xlabel('t')
plt.legend(loc='lower right')
plt.show()
plt.figure(dpi=100)
plt.plot(t, G3, 'g', label='G3')
plt.xlabel('t')
plt.legend(loc='lower right')
plt.show()
```

Appendix II

```
import numpy as np
from numpy import linalg as LA
a=[[0,1,1,1,1,1,0,0,0,0,0,0,0], [1,0,0,0,0,0,0,0,0,0,0,0,0],
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[1,0,0,0,0,0,0,0,0,0,0,0], [1,0,0,0,0,0,0,0,0,0,0,0],
[0,0,0,0,0,0,0,0,1,0,0,0,0], [0,0,0,1,0,0,0,0,0,0,0],
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