

## HOMEWORK 3 - CS323

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### Problem 1.

**a.**

$$\|Tx\| \leq \|T\| \|x\|$$

$$\text{Since, } \|T^k\| = \|T * T * T * \dots T_k\|$$

$$\text{and we know for some A and B, } \|A * B\| \leq \|A\| * \|B\|$$

$$\text{then } \|T^k\| = \|T * T * T * \dots T_k\| \leq \|T\| * \|T\| * \|T\| * \dots \|T_k\|$$

$$\|T^k x\| \leq \|T^k\| \|x\| \leq \|T\| * \|T\| * \|T\| * \dots \|T_k\| = \|T\|^k \|x\|$$

$$\|T^k x\| \leq \|T\|^k \|x\|$$

**b.**

$$A = D - L - U$$

$$\text{so, } (D - L - U)x = b$$

$$\text{then, } Dx - Lx - Ux = b$$

$$Dx = (L + U)x + b$$

$$Dx^{k+1} = (L + U)x^k + b$$

$$x^{k+1} = D^{-1}(L + U)x^k + b$$

$$\text{with } T = D^{-1}(L + U), \text{ then } \implies x^{k+1} = Tx^k + c$$

**c.**

$$x\star = Tx\star + c$$

$$e^k = x^k - x\star$$

$$\text{then, } e^{k+1} = (Tx^k + c) - (Tx\star + c)$$

$$e^{k+1} = Tx^k - Tx\star$$

$$e^{k+1} = T(x^k - x\star)$$

$$\implies e^{k+1} = Te^k, \text{ where } e^k = x^k - x\star$$

### Problem 2.

The error is reduced as the degree of each polynomial increases. The polynomial of degree  $n = 5$  captures the trend of the data best.

The polynomials of degree  $n = 0, 1, \dots, 5$  :

5.583

$$1.706 x + 1.319$$

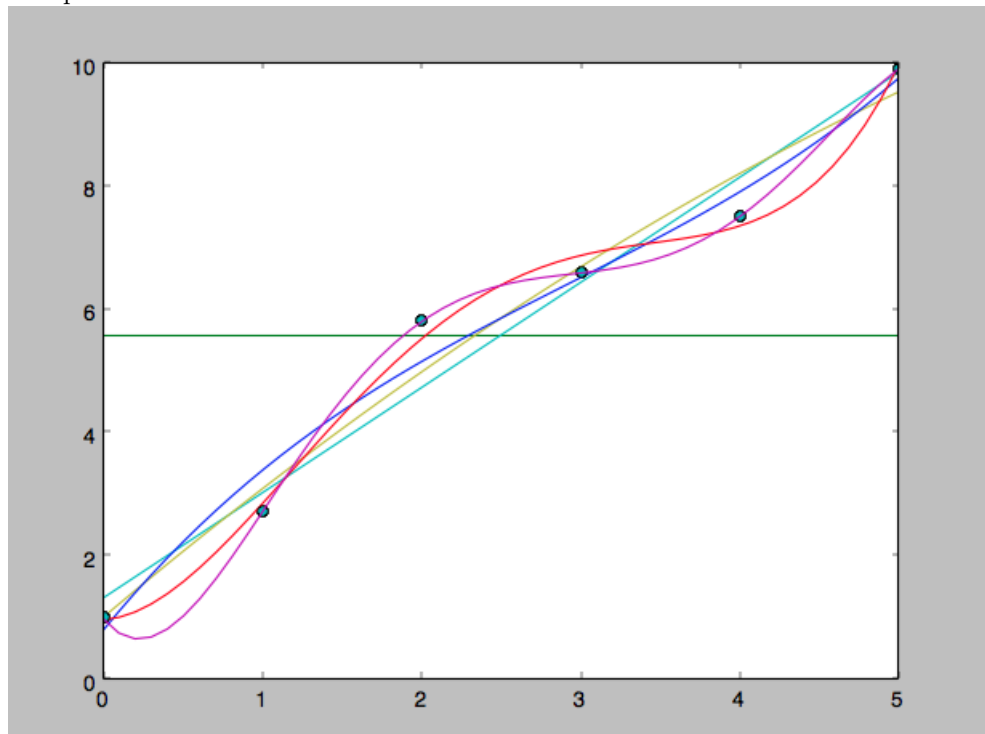
$$-0.09464 x^2 + 2.179 x + 1.004$$

$$0.0713 x^3 - 0.6294 x^2 + 3.156 x + 0.7897$$

$$0.1062 x^4 - 0.9912 x^3 + 2.634 x^2 + 0.12 x + 0.9718$$

$$-0.05917 x^5 + 0.8458 x^4 - 4.213 x^3 + 8.304 x^2 - 3.178 x + 1$$

The plots:



Code:

```

4
5 import matplotlib.pyplot as plt
6 from matplotlib import pylab
7 import numpy as np
8
9 points = np.array([(0.0, 1.), (1., 2.7), (2., 5.8), (3, 6.6), (4.,7.5), (5.,9.9)])
10 x = points[:,0]
11 y = points[:,1]
12
13 # calculate polynomials
14 z0 = np.polyfit(x, y, 0)
15 z1 = np.polyfit(x, y, 1)
16 z2 = np.polyfit(x, y, 2)
17 z3 = np.polyfit(x, y, 3)
18 z4 = np.polyfit(x, y, 4)
19 z5 = np.polyfit(x, y, 5)
20 f0 = np.poly1d(z0)
21 f1 = np.poly1d(z1)
22 f2 = np.poly1d(z2)
23 f3 = np.poly1d(z3)
24 f4 = np.poly1d(z4)
25 f5 = np.poly1d(z5)
26
27 #print polynomials
28 print f0, "\n\n", f1, "\n\n", f2, "\n\n", f3, "\n\n", f4, "\n\n", f5, "\n"
29
30 #new x's and y's
31 x1_new = np.linspace(x[0], x[-1], 50)
32 y0_new = f0(x1_new)
33 plt.plot(x,y,'o', x1_new, y0_new)
34
35 y1_new = f1(x1_new)
36 plt.plot(x,y,'o', x1_new, y1_new)
37
38 y2_new = f2(x1_new)
39 plt.plot(x,y,'o', x1_new, y2_new)
40
41 y3_new = f3(x1_new)
42 plt.plot(x,y,'o', x1_new, y3_new)
43
44 y4_new = f4(x1_new)
45 plt.plot(x,y,'o', x1_new, y4_new)
46
47 y5_new = f5(x1_new)
48 plt.plot(x,y,'o', x1_new, y5_new)
49
50 plt.show()

```

### Problem 3.

a.

$x_1 = 1, x_2 = 1$

b.

$x_1 = 7.00888731, x_2 = -8.39566299$

c.

The results are very different given the slightly modified b vector. A possible reason for this is because the matrix is ill-conditioned, meaning the condition number is very large. The condition number and resulting x values can be seen in the output below.

```
L1 cond(A^T*A): 1631924.81894
L2 cond(A^T*A): 1204591.14638

For b = [ 0.26  0.28  3.31]    x = [ 1.  1.]
For b = [ 0.27  0.25  3.33]    x = [ 7.00888731 -8.39566299]
```

Code:

```
import numpy as np

A = np.array([[0.16,0.1],[0.17,0.11],[2.02,1.29]])
b = np.array([0.26,0.28,3.31])
b2 = np.array([0.27,0.25,3.33])

#newA = A^T*A
# ??? ValueError: operands could not be broadcast together with shapes (2,2) and (3,2)

newA = np.dot(np.transpose(A) , A)
print "\nL1 cond(A^T*A): ", np.linalg.cond(newA,1)
print "L2 cond(A^T*A): ", np.linalg.cond(newA)

print ""
newb = np.dot(np.transpose(A) , b)
x = np.linalg.solve(newA,newb)
print "For b = ",b,"\t x = ", x

newb2 = np.dot(np.transpose(A) , b2)
x2 = np.linalg.solve(newA,newb2)
print "For b = ",b2,"\t x = ", x2
```

#### Problem 4.

a.

In order to show that the matrix is singular, we will show that its determinant is equal to 0.

$$A = \begin{bmatrix} 0.1 & 0.2 & 0.3 \\ 0.4 & 0.5 & 0.6 \\ 0.7 & 0.8 & 0.9 \end{bmatrix}$$

$$\Rightarrow \det(A) = \begin{vmatrix} 0.1 & 0.2 & 0.3 \\ 0.4 & 0.5 & 0.6 \\ 0.7 & 0.8 & 0.9 \end{vmatrix} = 0.1 \begin{vmatrix} 0.5 & 0.6 \\ 0.8 & 0.9 \end{vmatrix} - 0.2 \begin{vmatrix} 0.4 & 0.6 \\ 0.7 & 0.9 \end{vmatrix} + 0.3 \begin{vmatrix} 0.4 & 0.5 \\ 0.7 & 0.8 \end{vmatrix}$$

$$= 0.1(0.45 - 0.48) - 0.2(0.36 - 0.42) + 0.3(0.32 - 0.35)$$

$$= 0$$

$$\text{If } b = \begin{bmatrix} 0.1 \\ 0.3 \\ 0.5 \end{bmatrix}, \text{ the set of solutions to } Ax = b \text{ is } x = \left\{ x_1 - x_3 = \frac{1}{3}, \quad x_2 + 2x_3 = \frac{1}{3} \right\}$$

This system is dependent so there are infinitely many solutions.

b.

$$\begin{bmatrix} \frac{1}{10} & \frac{2}{10} & \frac{3}{10} \\ \frac{4}{10} & \frac{5}{10} & \frac{6}{10} \\ \frac{7}{10} & \frac{8}{10} & \frac{9}{10} \end{bmatrix} \Rightarrow \begin{bmatrix} \frac{7}{10} & \frac{8}{10} & \frac{9}{10} \\ \frac{4}{10} & \frac{5}{10} & \frac{6}{10} \\ \frac{1}{10} & \frac{2}{10} & \frac{3}{10} \end{bmatrix} \Rightarrow \begin{bmatrix} \frac{7}{10} & \frac{8}{10} & \frac{9}{10} \\ 0 & \frac{-3}{40} & \frac{-3}{20} \\ 0 & \frac{-3}{5} & \frac{-6}{5} \end{bmatrix} \Rightarrow \begin{bmatrix} \frac{7}{10} & \frac{8}{10} & \frac{9}{10} \\ 0 & \frac{-3}{40} & \frac{-3}{20} \\ 0 & 0 & 0 \end{bmatrix} \Rightarrow$$

$$\begin{bmatrix} \frac{7}{10} & \frac{8}{10} & \frac{9}{10} \\ 0 & \frac{-3}{5} & \frac{-6}{5} \\ 0 & 0 & 0 \end{bmatrix}$$

At this point in the process, when the entire row becomes zeros, the gaussian elimination fails.

c.

The computed solution becomes  $\begin{bmatrix} 0.08333333 & 0.83333333 & -0.25 \end{bmatrix}$ . Although this is a solution to the system, it is only one of an infinite amount that we found in part a.

The estimated condition number of this matrix is  $8.64691128455e+16$