

# Numerical Method Homework3

Xiangbo Wang

May 10, 2011

## 1 Problem1

From the definition of norm of matrix,  $\|A\|$  bounds the "amplifying power" of the matrix:  $\|Ax\| \leq \|A\|\|x\|$ . And since  $b = Ax$  and  $\delta x = A^{-1}\delta b$ , we could obtain that

$$\|b\| \leq \|A\|\|x\| \quad (1)$$

$$\|\delta x\| \leq \|A^{-1}\|\|\delta b\| \quad (2)$$

Then from equation (1), we could obtain that

$$\frac{1}{\|x\|} \leq \|A\| \frac{1}{\|b\|} \quad (3)$$

Therefore from both equation (2) and equation (3), we get what we need

$$\frac{\|\delta x\|}{\|x\|} \leq \|A\|\|A^{-1}\| \frac{\|\delta b\|}{\|b\|} \quad (4)$$

## 2 Problem2

### 2.1 Norm of an orthonormal matrix

The norm of matrix  $M$  could be obtained by following equation

$$\|M\| = \sqrt{\text{largest eigenvalue of } M^*M} \quad (5)$$

*where  $M^*M = M^T M$*

Since matrix  $M$  is an orthonormal matrix, it has the properties of  $M^T = M^{-1}$  and  $M^{-1}M = I$ . Then  $M^*M = M^T M = M^{-1}M = I$ . Because the eigenvalue of  $I$  is 1, therefore the norm of an orthonormal matrix  $M$  is 1 or  $-1$ .

## 2.2 Determinant of an orthonormal matrix

Let's assume that the orthogonal matrix is  $M$ . And its transpose is  $M^T$ .

According to the properties of transpose of a matrix, we obtain that  $\det(M) = \det(M^T)$ . Then from that equation, we get that  $\det(M)^2 = \det(M)\det(M) = \det(M^T)\det(M) = \det(MM^T)$ . As  $M$  is an orthogonal matrix, it has the property that  $M^T = M^{-1}$  which entails that  $MM^T = I$ . Based on all we get, we finally obtain that  $\det(MM^T) = \det(I) = 1$ .

Therefore the determinant of an orthonormal matrix is 1 or  $-1$ .

## 3 Problem3

To show that any two eigenvectors of a symmetric matrix are orthogonal, we must show that the dot product of any two eigenvectors is zero. Let's assume that  $v_1$  and  $v_2$  are eigenvectors corresponding to distinct eigenvalues  $\lambda_1$  and  $\lambda_2$  of matrix  $A$ . So we have  $Av_1 = \lambda_1 v_1$  and  $Av_2 = \lambda_2 v_2$ . Then we can obtain that

$$\lambda_1(v_1 \cdot v_2) = (\lambda_1 v_1) \cdot v_2 = (Av_1) \cdot v_2 \quad (6)$$

According to the properties of dot product of matrix, we have

$$(Av_1) \cdot v_2 = (Av_1)^T v_2 = (v_1^T A^T) v_2 = v_1^T (A^T v_2) = v_1 \cdot (A^T v_2) \quad (7)$$

Since  $A$  is a symmetric matrix, which has the property that  $A = A^T$ , then equation(2) will be

$$v_1 \cdot (A^T v_2) = v_1 \cdot (Av_2) = v_1 \cdot (\lambda_2 v_2) = \lambda_2(v_1 \cdot v_2) \quad (8)$$

Hence

$$\lambda_1(v_1 \cdot v_2) = \lambda_2(v_1 \cdot v_2) \Rightarrow (\lambda_1 - \lambda_2)(v_1 \cdot v_2) = 0 \quad (9)$$

Since  $\lambda_1$  and  $\lambda_2$  are distinct, we can conclude that  $v_1 \cdot v_2 = 0$ , which means  $v_1$  and  $v_2$  are orthogonal. Therefore the eigenvectors of a symmetric matrix are orthogonal.

## 4 Problem4

Let's assume there exists such  $y$  minimize the equation like

$$\frac{1}{2}y^T Ay - b^T y + c = a \text{ where } a \text{ is the minimum value} \quad (10)$$

And if  $Ay = b$  holds, from equation (10) we should have

$$\frac{1}{2}y^T Ay - (Ay)^T y + c = \frac{1}{2}y^T Ay - y^T A^T y + c = a \quad (11)$$

Since  $A$  is symmetric, then it has the property of  $A = A^T$ . Then equation (11) will be

$$\frac{1}{2}y^T Ay - y^T Ay + c = -\frac{1}{2}y^T Ay + c = a \quad (12)$$

Since  $A$  is positive definite,  $y^T Ay$  is always positive, which means  $-\frac{1}{2}y^T Ay$  is always negative. Therefore equation (12) should hold for certain minimized  $a$ , and such  $y$  must exist which holds the two equations.

## 5 Problem5

I use *p5.cpp* to implement the solver. Basic idea is that do *Gaussian Elimination* on the matrix  $A$  first, then solve the problem according the eliminated matrix. Since we know that the matrix is a quint-diagonal matrix, I decrease the operations and memory usage by just operating on the non-zero elements.

*ps.* please use gcc to compile the file.

## 6 Problem6

I use *p6\_pbc.m* and *p6\_zbc.m* to implement *periodic boundary conditions* and *zero-padded boundary conditions* respectively. From the results I obtain, both methods work fine, but *pbc* is better.

Both methods have a determinant of 0. Determinant shows the measure of volume is multiplied under the transformation, so the result shows that both of them work good. Also the eigenvectors/eigenvalues look good and similar between them. However, *zbc* has an *Inf* condition number while *pbc* always has a really small one. Condition numbers measure the asymptotically worst case of how much the function can change in proportion to small changes in the argument. And one with a low condition number is said to be well-conditioned, while one with a high condition number is said to be ill-conditioned. In conclusion, *pbc* works better than *zbc*.

*ps.* The input of both programs must follow the format as *[1 2 3 4 5 6]*

## 7 Problem7

I use *p7.m* to implement the *pair difference*. The results shows that the matrix has a determinant of 1 which is relevantly high compared with the previous two. Besides, it has a fair condition number which shows that it works fine.

*ps.* The input of both programs must follow the format as *[1 2 3 4 5 6]*

## 8 Problem8

### 8.1 LU

I use *Crout's Method* to solve this problem. Since the upper and lower triangle are all zeros, the L and U should be like

$$L = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ l_{21} & 1 & 0 & 0 & 0 \\ 0 & l_{32} & 1 & 0 & 0 \\ 0 & 0 & l_{43} & 1 & 0 \\ 0 & 0 & 0 & l_{54} & 1 \end{pmatrix} \quad (13)$$

$$U = \begin{pmatrix} u_{11} & u_{12} & 0 & 0 & 0 \\ 0 & u_{22} & u_{23} & 0 & 0 \\ 0 & 0 & u_{33} & u_{34} & 0 \\ 0 & 0 & 0 & u_{44} & u_{45} \\ 0 & 0 & 0 & 0 & u_{55} \end{pmatrix} \quad (14)$$

Then I multiple these two matrices and get the following equations

$$\begin{aligned}
1 \times u_{11} &= 4 \\
1 \times u_{12} &= 1 \\
l_{21} \times u_{11} &= 1 \\
l_{21} \times u_{12} + 1 \times u_{22} &= 4 \\
1 \times u_{23} &= 1 \\
l_{32} \times u_{22} &= 1 \\
l_{32} \times u_{23} + 1 \times u_{33} &= 4 \\
1 \times u_{34} &= 1 \\
l_{43} \times u_{33} &= 1 \\
l_{43} \times u_{34} + 1 \times u_{44} &= 4 \\
1 \times u_{45} &= 1 \\
l_{54} \times u_{44} &= 1 \\
l_{54} \times u_{45} + 1 \times u_{55} &= 4
\end{aligned}$$

From these equations, I obtain L and U which are

$$L = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ \frac{1}{4} & 1 & 0 & 0 & 0 \\ 0 & \frac{4}{15} & 1 & 0 & 0 \\ 0 & 0 & \frac{15}{56} & 1 & 0 \\ 0 & 0 & 0 & \frac{56}{209} & 1 \end{pmatrix} \quad (15)$$

$$U = \begin{pmatrix} 4 & 1 & 0 & 0 & 0 \\ 0 & \frac{15}{4} & 1 & 0 & 0 \\ 0 & 0 & \frac{56}{15} & 1 & 0 \\ 0 & 0 & 0 & \frac{209}{56} & 1 \\ 0 & 0 & 0 & 0 & \frac{780}{209} \end{pmatrix} \quad (16)$$

## 8.2 Symmetric

A matrix is symmetric if its transpose and itself are the same ( $M = M^T$ ). In this question,  $A$ 's transpose  $A^T$  equals to  $A$ , which shows that  $A$  is symmetric.

### 8.3 Positive Definite

Since A is a real symmetric matrix, it is positive definite if  $z^T M z > 0$  for all non-zero vectors  $z$ . Therefore I obtain the equation

$$\begin{bmatrix} z_1 & z_2 & z_3 & z_4 & z_5 \end{bmatrix} \begin{bmatrix} 4 & 1 & 0 & 0 & 0 \\ 1 & 4 & 1 & 0 & 0 \\ 0 & 1 & 4 & 1 & 0 \\ 0 & 0 & 1 & 4 & 1 \\ 0 & 0 & 0 & 1 & 4 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \\ z_5 \end{bmatrix} \quad (17)$$

After multiplying these three matrices, I obtain the following equation

$$\begin{aligned} 4z_1^2 + 2z_1z_2 + 4z_2^2 + 2z_2z_3 + 4z_3^2 + 2z_3z_4 + 4z_4^2 + 2z_4z_5 + 4z_5^2 \\ = 3z_1^2 + (z_1 + z_2)^2 + 2z_2^2 + (z_2 + z_3)^2 + 2z_3^2 + \\ (z_3 + z_4)^2 + 2z_4^2 + (z_4 + z_5)^2 + 3z_5^2 > 0 \end{aligned} \quad (18)$$

Thus A is positive definite.

### 8.4 Invertible

Matrix A is invertible if there exists a matrix B such that  $AB = BA = I_n$ . Then for the matrix in this question, I find a matrix B which meets that  $AB = BA = I_n$ .

$$B = \begin{pmatrix} 0.2679 & -0.0718 & 0.0192 & -0.0051 & 0.0013 \\ -0.0718 & 0.2872 & -0.0769 & 0.0205 & -0.0051 \\ 0.0192 & -0.0769 & 0.2885 & -0.0769 & 0.0192 \\ -0.0051 & 0.0205 & -0.0769 & 0.2872 & -0.0718 \\ 0.0013 & -0.0051 & 0.0192 & -0.0718 & 0.2679 \end{pmatrix} \quad (19)$$

Therefore A is invertible.

### 8.5 Largest Eigenvalue

I use *p8.e.m* computing the largest eigenvalue by *Power Iteration Method*. And the result is 5.7312.

### 8.6 All Eigenvalue

I use *p8.f.m* computing all of the eigenvalues of A by *QR Method*. And the five eigenvalues are 5.7321, 5, 4, 3, 2.2679.

## 9 Problem9