

# DS284: Numerical Linear Algebra

## Assignment 4

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**Office Hours:** Sat 2 PM to 4 PM

Total: 100 Marks

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### Problem 1

[20 marks]

In this problem you will test different algorithms for the least squares problem for approximating the polynomial function  $f(t) = \sin(10t)$  for  $t \in [0, 1]$ . To set it up follow the example discussed in the class. Choose the number of points  $m = 100$  and  $n = 15$  (14th degree polynomial). Determine its least square solutions with the following methods.

- (a) Use QR Factorization with your implementation of Modified Gram Schmidt. You should write your own back substitution code for solving the resulting triangular system.
- (b) Using QR Factorization with your implementation of Householder factorization.
- (c) Using SVD (Computed with any inbuilt libraries in MATLAB/Python/Octave)
- (d) Using normal equations, you can use backslash command in MATLAB to solve this system.

Accept the MATLAB/Octave/Python least squares solution (given by backslash “\” in MATLAB) as the truth. Display and plot the approximation given by this “true” solution and compare it with  $f(t)$ . Compare with the solution given by 4 methods described above. Explain the results.

### Problem 2

[10 marks]

Given below is the matrix  $\mathbf{A}$  and vector  $\mathbf{b}$

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 1 & 2 & 3 \\ 2 & 8 & 2 \end{bmatrix} \quad \mathbf{b} = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$$

Find the least square solution for the above. Now do the same for the below matrix also.

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 1 & 2 & 3 \\ 5 & 7 & 9 \end{bmatrix}$$

Were you able to find the least square solution? If not explain why? Must there be a restriction on  $\mathbf{A}$  for a least square solution to exist or will it always exist?

**Problem 3****[10 marks]**

Gaussian Elimination allows us to compute determinant of a square matrix. Recall the following points about the determinant.

- Swapping 2 rows multiplies the determinant by -1.
- Multiplying a row by a non-zero scalar multiplies the determinant by the same scalar.
- Adding one row, which is a scalar multiple of the other does not change the determinant.

Now describe the procedure how Gaussian Elimination with partial pivoting can be used to find the determinant of a general square matrix. Also comment on the number of floating point operations in this procedure!

**Problem 4****[15 marks]**

Let  $\mathbf{A}$  be a matrix defined in MATLAB as:

```
A = rand(N)
A = A - diag(diag(A)) + diag(0.001 * ones(N,1))
```

Compute LU Decomposition of  $\mathbf{A}$  with and without partial pivoting. Plot  $\|\mathbf{LU} - \mathbf{A}\|_F$  versus  $N$  for  $N = 5, 6, 7, \dots, 20$ . For LU Decomposition with partial pivoting, use built in LU function. For LU Decomposition without pivoting, write your own function.

**Problem 5****[15 marks]**

- Let  $\mathbf{A}$  be a non-singular square matrix and let  $\mathbf{A} = \mathbf{QR}$  be its QR factorization. Let also  $\mathbf{A}^T \mathbf{A} = \mathbf{U}^T \mathbf{U}$  be the Cholesky factorization of  $\mathbf{A}^T \mathbf{A}$ . Can you conclude that  $\mathbf{R} = \mathbf{U}$ ? If yes, prove it; if not, why not?
- Recall that by  $\mathbf{A} \in \mathbb{R}^{m \times m}$ , being symmetric and strictly positive definite, we mean  $\mathbf{A} = \mathbf{A}^T$  and  $\forall \mathbf{x} \in \mathbb{R}^n, \mathbf{x} \neq 0$ , we have  $\mathbf{x}^T \mathbf{A} \mathbf{x} > 0$ . A symmetric matrix  $\mathbf{A} \in \mathbb{R}^{m \times m}$  is positive semi-definite if  $\forall \mathbf{x} \in \mathbb{R}^n, \mathbf{x} \neq 0$ , we have  $\mathbf{x}^T \mathbf{A} \mathbf{x} \geq 0$

If  $\{\phi_i(x)\}_{i=1 \dots m}$  denote  $m$  linearly independent basis functions defined over  $[-1, 1]$  in an  $m$ -dimensional vector space then show that the matrix  $\mathbf{M} = \int_{-1}^1 \phi_i(x) \phi_j(x) dx$  for  $i, j = 1, 2 \dots m$  is a symmetric positive definite matrix.

Similarly show that the matrix  $\mathbf{K} = \int_{-1}^1 \frac{d\phi_i(x)}{dx} \frac{d\phi_j(x)}{dx} dx$  for  $i, j = 1, 2 \dots m$  is a symmetric positive semi-definite matrix.

- Show that a matrix  $\mathbf{A} \in \mathbb{R}^{n \times n}$  is symmetric, strictly positive definite if and only if there exists a matrix  $\mathbf{B} \in \mathbb{R}^{m \times n}$  of rank  $n$ , where  $n \leq m$ , such that  $\mathbf{A} = \mathbf{B}^T \mathbf{B}$ . Assuming that  $\mathbf{A}$  is of this form, is there a unique such  $\mathbf{B}$ ?

**Problem 6****[12 marks]**

For each of the following statements prove that it is true or give an example to show that it is false. Assume  $\mathbf{A} \in \mathbb{C}^{m \times m}$  unless otherwise indicated.

- If  $\lambda$  is an eigenvalue of  $\mathbf{A}$  and  $\mu \in \mathbb{C}$ , then  $\lambda - \mu$  is an eigenvalue of  $\mathbf{A} - \mu \mathbf{I}$ .
- If  $\mathbf{A}$  is real and  $\lambda$  is an eigenvalue of  $\mathbf{A}$  then so is  $-\lambda$ .
- If  $\mathbf{A}$  is real and  $\lambda$  is an eigenvalue of  $\mathbf{A}$ , then so is  $\lambda^*$ . ( $\lambda^*$  is the complex conjugate of  $\lambda$ ).
- If  $\lambda$  is an eigenvalue of  $\mathbf{A}$  and  $\mathbf{A}$  is non-singular, then  $\lambda^{-1}$  is the eigenvalue of  $\mathbf{A}^{-1}$ .
- If all the eigenvalues of  $\mathbf{A}$  are zero, then  $\mathbf{A} = 0$ .

- (f) If  $\mathbf{A}$  is diagonalizable and all its eigenvalues are equal, then  $\mathbf{A}$  is diagonal.

## Problem 7

[18 marks]

Let  $\mathbf{A} \in \mathbb{C}^{n \times n}$  with entries  $a_{ij}$  for  $i, j = 1, 2, \dots, n$  and define the closed disks  $D(a_{ii}, r_i)$  centered at the diagonal entries  $a_{ii}$  of  $\mathbf{A}$  of radius  $r_i = \sum_{j=1}^n (1 - \delta_{ij}) |a_{ij}|$  for  $i = 1, 2, \dots, n$ . Note that  $\delta_{ij}$  represents Kronecker delta i.e.  $\delta_{ij} = 1$  if  $i = j$  and  $\delta_{ij} = 0$  if  $i \neq j$ . The above disks are called Greshgorin's disks.

- Prove that every eigenvalue of  $\mathbf{A}$  lies in a Greshgorin disk.  
(*Hint:* Let  $\lambda$  be any eigenvalue of  $\mathbf{A}$  and  $\mathbf{x}$  be the corresponding eigenvector with largest entry 1.)
- Suppose that  $\mathbf{A}$  is diagonally dominant i.e.  $|a_{ii}| > \sum_{j=1}^n (1 - \delta_{ij}) |a_{ij}|$  for all  $i = 1, 2, \dots, n$ . Prove that  $\mathbf{A}$  is invertible.
- Give estimates based on (a), for the eigenvalues of:

$$\mathbf{A} = \begin{bmatrix} 8 & 2 & 0 \\ 1 & 4 & \epsilon \\ 0 & \epsilon & 1 \end{bmatrix} \quad \text{where } |\epsilon| < 1$$