

**STAT 309: MATHEMATICAL COMPUTATIONS I**  
**FALL 2011**  
**PROBLEM SET 3**

MATLAB is recommended for Problems 1 and 2 but you're of course free to use any other programs.

1. Let  $A = [a_{ij}]$  be an  $n \times n$  matrix with entries

$$a_{ij} = \begin{cases} n + 1 - \max(i, j) & i \leq j + 1, \\ 0 & i > j + 1. \end{cases}$$

This is an example of an *upper Hessenberg* matrix: it is upper triangular except that the entries on the subdiagonal  $a_{i,i+1}$  may also be non-zero. For  $n = 12$  and  $n = 25$ , do the following:

- (a) Compute  $\|A\|_\infty$  and  $\|A\|_1$ .
- (b) Compute  $\rho(A)$  and  $\|A\|_2$ .
- (c) Using Gerschgorin's theorem, describe the domain that contains all of the eigenvalues.
- (d) Compute all of the eigenvalues and singular values of  $A$ . How many of the eigenvalues are real and how many are complex?

2. Let  $A = [a_{ij}]$  be a  $32 \times 32$  matrix defined by

$$a_{ij} := \begin{cases} 1 & \text{if } j = i, \\ i - 11 & \text{if } j = i + 1, \\ 0 & \text{otherwise.} \end{cases} \quad (2.1)$$

In other words,  $A$  is a bidiagonal matrix with 1's on its diagonal,  $-10, -9, \dots, -1, 1, \dots, 21$  on its superdiagonal, and 0's everywhere else.

- (a) Construct the Gerschgorin disks for  $A$ .
- (b) Let  $\varepsilon > 0$ . Construct a diagonal matrix  $D$  so that  $DAD^{-1}$  is bidiagonal with 1's on its diagonal,  $\varepsilon$ 's on its superdiagonal, and 0's everywhere else, ie.

$$DAD^{-1} = \begin{bmatrix} 1 & \varepsilon & & & \\ & 1 & \varepsilon & & \\ & & 1 & \ddots & \\ & & & \ddots & \varepsilon \\ & & & & 1 \end{bmatrix}. \quad (2.2)$$

What are the Gerschgorin disks for  $DAD^{-1}$ ?

- (c) Give an algorithm to reduce  $A$  in (2.1) to the form in (2.2) with  $\varepsilon = 10^{-4}$ . How stable is your algorithm?

3. Let  $A, B \in \mathbb{R}^{m \times n}$  with  $n \leq m$ . In the lectures, we claim that the solution  $Q \in O(m)$  to

$$\min_{Q^T Q = I} \|A - BQ\|_F$$

is given by  $Q = UV^T$  where  $B^T A = U\Sigma V^T$  is its singular value decomposition. Here we will prove it.

(a) Show that

$$\|A - BQ\|_F^2 = \text{tr}(A^\top A) + \text{tr}(B^\top B) - 2\text{tr}(Q^\top B^\top A)$$

and deduce that the minimization problem is equivalent to

$$\max_{Q^\top Q=I} \text{tr}(Q^\top B^\top A).$$

(b) Show that

$$\text{tr}(Q^\top B^\top A) \leq \sum_{i=1}^n \sigma_i(B^\top A)$$

for any  $Q \in O(n)$ . When is the upper bound attained?

(c) Show that

$$\min_{Q^\top Q=I} \|A - BQ\|_F^2 = \sum_{i=1}^m (\sigma_i(A)^2 - 2\sigma_i(B^\top A) + \sigma_i(B)^2).$$

(d) Suppose  $A$  has full rank. Describe how you would find a symmetric matrix  $X \in \mathbb{R}^{n \times n}$  that solves

$$\min_{X=X^\top} \|AX - B\|_F.$$

(Hint: Consider the singular value decomposition of  $A$ ).

4. Let  $\mathbf{x} \in \mathbb{C}^m$ ,  $\mathbf{y} \in \mathbb{C}^n$ , and  $A = \mathbf{x}\mathbf{y}^* \in \mathbb{C}^{m \times n}$ .

(a) Show that  $\text{rank}(A) = 1$  iff  $\mathbf{x}$  and  $\mathbf{y}$  are both non-zero. Such a matrix is usually called a rank-1 matrix.

(b) Show that

$$\|A\|_F = \|A\|_2 = \|\mathbf{x}\|_2 \|\mathbf{y}\|_2$$

and that

$$\|A\|_\infty \leq \|\mathbf{x}\|_\infty \|\mathbf{y}\|_1.$$

What can you say about  $\|A\|_1$ ?

(c) Let  $\mathbf{x}_1, \dots, \mathbf{x}_r \in \mathbb{C}^m$  be linearly independent and  $\mathbf{y}_1, \dots, \mathbf{y}_r \in \mathbb{C}^n$  be linearly independent. Let

$$A = \mathbf{x}_1\mathbf{y}_1^* + \dots + \mathbf{x}_r\mathbf{y}_r^*.$$

Show that  $\text{rank}(A) = r$ . Show that this is not necessarily true if we drop either of the linear independence conditions.

(d) Given any  $0 \neq A \in \mathbb{C}^{m \times n}$ , show that

$$\text{rank}(A) = \min\{r \in \mathbb{N} \mid A = \sum_{i=1}^r \mathbf{x}_i\mathbf{y}_i^*\}.$$

In other words, the rank of a matrix is the smallest  $r$  so that it may be expressed as a sum of  $r$  rank-1 matrices.

(e) Show that more generally,

$$\|A\|_F \leq \sqrt{\text{rank}(A)} \|A\|_2.$$

Note that  $\rho \text{rank}(A) = \|A\|_F^2 / \|A\|_2^2$  is the ‘computer scientist’s numerical rank,’ one of the three notions of numerical ranks that we discussed.

5. Let  $A \in \mathbb{R}^{m \times n}$  and  $\mathbf{b} \in \mathbb{R}^m$ . We will discuss a variant of  $A\mathbf{x} \approx \mathbf{b}$  where the error occurs only in  $A$ . Note that in ordinary least squares we assume that the error occurs only in  $\mathbf{b}$  while in total least squares we assume that it occurs in both  $A$  and  $\mathbf{b}$ .

(a) Show that if  $0 \neq \mathbf{x} \in \mathbb{R}^m$ , then

$$\left\| A \left( I - \frac{\mathbf{x}\mathbf{x}^\top}{\mathbf{x}^\top \mathbf{x}} \right) \right\|_F^2 = \|A\|_F^2 - \frac{\|A\mathbf{x}\|_2^2}{\mathbf{x}^\top \mathbf{x}}.$$

(b) Show that the matrix

$$E = \frac{(\mathbf{b} - A\mathbf{x})\mathbf{x}^\top}{\mathbf{x}^\top \mathbf{x}} \in \mathbb{R}^{m \times n}$$

has the smallest 2-norm of all  $m \times n$  matrices  $E$  that satisfy

$$(A + E)\mathbf{x} = \mathbf{b}.$$

(c) What are the solutions of

$$\min_{(A+E)\mathbf{x}=\mathbf{b}} \|E\|_2 \quad \text{and} \quad \min_{(A+E)\mathbf{x}=\mathbf{b}} \|E\|_F?$$

(d) Given  $\mathbf{a} \in \mathbb{R}^n$ ,  $\mathbf{b} \in \mathbb{R}^m$ , and  $\delta > 0$ . Show how to solve the problems

$$\min_{\|E\|_F \leq \delta} \|E\mathbf{a} - \mathbf{b}\|_2 \quad \text{and} \quad \max_{\|E\|_F \leq \delta} \|E\mathbf{a} - \mathbf{b}\|_2$$

over all  $E \in \mathbb{R}^{m \times n}$ .

**6.** Let  $\mathbf{u} \in \mathbb{C}^n$ ,  $\mathbf{u} \neq \mathbf{0}$ . A *Householder* matrix  $H_{\mathbf{u}} \in \mathbb{C}^{n \times n}$  is defined by

$$H_{\mathbf{u}} = I - \frac{2\mathbf{u}\mathbf{u}^*}{\|\mathbf{u}\|_2^2}.$$

(a) Show that  $H_{\mathbf{u}}$  is both Hermitian and unitary.

(b) Show that for any  $\alpha \in \mathbb{C}$ ,  $\alpha \neq 0$ ,

$$H_{\alpha\mathbf{u}} = H_{\mathbf{u}}.$$

In other words,  $H_{\mathbf{u}}$  only depends on the ‘direction’ of  $\mathbf{u}$  and not on its ‘magnitude’.

(c) In general, given a matrix  $M \in \mathbb{C}^{n \times n}$  and a vector  $\mathbf{x} \in \mathbb{C}^n$ , computing the matrix-vector product  $M\mathbf{x}$  requires  $n$  inner products — one for each row of  $M$  with  $\mathbf{x}$ . Show that  $H_{\mathbf{u}}\mathbf{x}$  can be computed using only two inner products.

(d) Given  $\mathbf{a}, \mathbf{b} \in \mathbb{C}^n$  where  $\mathbf{a} \neq e^{i\theta}\mathbf{b}$  for any  $\theta \in [0, 2\pi)$  and  $\|\mathbf{a}\|_2 = \|\mathbf{b}\|_2$ . Find  $\mathbf{u} \in \mathbb{C}^n$ ,  $\mathbf{u} \neq \mathbf{0}$  such that

$$H_{\mathbf{u}}\mathbf{a} = \mathbf{b}.$$

(e) Show that  $\mathbf{u}$  is an eigenvector of  $H_{\mathbf{u}}$ . What is the corresponding eigenvalue?

(f) Show that every  $\mathbf{v} \in \text{span}\{\mathbf{u}\}^\perp$  (cf. orthogonal complement in Homework 2, Problem 5) is an eigenvector of  $H_{\mathbf{u}}$ . What are the corresponding eigenvalues? What is  $\dim(\text{span}\{\mathbf{u}\}^\perp)$ ?

(g) Find the eigenvalue decomposition of  $H_{\mathbf{u}}$ , i.e. find a unitary matrix  $U$  and a diagonal matrix  $\Lambda$  such that

$$H_{\mathbf{u}} = U\Lambda U^*.$$

(*Hint:* Gram-Schmidt algorithm).