

# Solutions to Math 104 Final Exam — Dec. 10, 2010

1. (20 points) Consider the following matrices that depend on a parameter  $\lambda \in \mathbb{C}$ :

$$A_\lambda = \begin{bmatrix} 1 & 0 \\ 2 & 0 \\ \lambda & -1 \end{bmatrix}$$

(a) For each  $\lambda$ , determine  $\|A_\lambda\|_F$  the Frobenius norm of  $A_\lambda$ .

(Note: technically  $\lambda \in \mathbb{C}$  which complicates things). The Frobenius norm of  $A_\lambda$  is the square-root of the sum of the squares (of the modulus) of the entries of  $A_F$ .

$$\|A\|_F = \sqrt{1^2 + 2^2 + (-1)^2 + |\lambda|^2} = \sqrt{6 + |\lambda|^2}$$

(b) For each  $\lambda$ , determine  $\|A_\lambda\|_2$  the induced 2-norm of  $A_\lambda$ .

In order to compute the two norm of  $A_\lambda$  we compute the largest singular value  $\sigma_1$  of  $A$ . This is because, by definition,  $\|A_\lambda\|_2 = \sigma_1$ . We have established in class that the singular values are the square roots of the eigenvalues of the matrix

$$B_\lambda = A_\lambda^* A_\lambda = \begin{bmatrix} 5 + |\lambda|^2 & -\bar{\lambda} \\ -\lambda & 1 \end{bmatrix}$$

The characteristic polynomial of this matrix is  $p_B(z) = \det(B - zI) = z^2 - (6 + |\lambda|^2)z + 5$  so by the quadratic formula the eigenvalues are

$$\alpha_\pm = \frac{6 + |\lambda|^2 \pm \sqrt{|\lambda|^4 + 12|\lambda|^2 + 16}}{2}$$

Hence, the largest singular value of  $A$  is

$$\sigma_1 = \sqrt{\frac{6 + |\lambda|^2 + \sqrt{|\lambda|^4 + 12|\lambda|^2 + 16}}{2}}$$

2. (20 points) Let  $A \in \mathbb{C}^{m \times m}$  be hermitian (i.e.  $A = A^*$ ). Let  $P \in \mathbb{C}^{m \times m}$  be the matrix representing orthogonal projection onto  $N(A)$ . Please show that  $X = A + P$  is invertible. (Hint: Think about the four fundamental subspaces of  $A$ ).

In order to show that  $X$  is invertible it suffices to show that  $N(X) = \{0\}$ . To that end we note that  $R(P) = N(A)$  by definition of  $P$  and that  $R(A) = R(A^*)$  as  $A$  is hermitian. Now suppose that  $\mathbf{x} \in N(X)$  so  $X\mathbf{x} = 0$  then  $(A + P)\mathbf{x} = 0$  so that  $\mathbf{y} = A\mathbf{x} = -P\mathbf{x}$ . In other words,  $\mathbf{y} \in R(A)$  and  $\mathbf{y} \in R(P)$  and so  $\mathbf{y} \in N(A) \cap R(A^*)$ . As  $N(A)$  and  $R(A^*)$  are complementary spaces this means that  $\mathbf{y} = 0$ . Hence,  $\mathbf{x} \in N(A)$  and so  $\mathbf{x} \in R(P)$ . But then  $\mathbf{x} = P\mathbf{x} = \mathbf{y} = 0$ .

3. (20 points) Let  $A \in \mathbb{R}^{4 \times 2}$  be the matrix

$$A = \begin{bmatrix} 1 & -7 \\ 1 & 1 \\ 3 & -7 \\ 5 & -9 \end{bmatrix}$$

(a) Find a reduced  $QR$  factorization of  $A$  i.e.  $A = \hat{Q}\hat{R}$ .

To begin the  $QR$  factorization algorithm we normalize the first column of  $A$  this yields

$$\mathbf{q}_1 = \frac{1}{6} \begin{bmatrix} 1 \\ 1 \\ 3 \\ 5 \end{bmatrix}$$

and  $r_{11} = 6$  the length of the first column. We next compute the inner product between  $\mathbf{q}_1$  and the second column to obtain  $r_{12} = -12$  then subtracting  $r_{12}\mathbf{q}_1$  from the second column yields

$$\hat{\mathbf{q}}_2 = \begin{bmatrix} -5 \\ 3 \\ -1 \\ 1 \end{bmatrix}$$

This has length 6 so that  $r_{22} = 6$  and

$$\mathbf{q}_2 = \frac{1}{6} \begin{bmatrix} -5 \\ 3 \\ -1 \\ +1 \end{bmatrix}$$

Hence

$$\hat{Q} = \frac{1}{6} \begin{bmatrix} 1 & -5 \\ 1 & 3 \\ 3 & -1 \\ 5 & 1 \end{bmatrix}$$

and

$$\hat{R} = \begin{bmatrix} 6 & -12 \\ 0 & 6 \end{bmatrix}$$

and the reduced  $QR$  factorization of  $A$  is  $A = \hat{Q}\hat{R}$ .

(b) Solve the following overdetermined system in the sense of least squares:

$$A\mathbf{x} = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}$$

By the  $QR$  factorization we know that orthogonal projection onto  $R(A)$  is given by  $\hat{Q}\hat{Q}^*$ . In particular, since  $A = \hat{Q}\hat{R}$  to solve  $A\mathbf{x} = \mathbf{b}$  in the sense of least squares, we solve

$$\hat{Q}\hat{R}\mathbf{x} = \hat{Q}\hat{Q}^*\mathbf{b}$$

that is to solve

$$\hat{R}\mathbf{x} = \hat{Q}^*\mathbf{b}.$$

In our situation, the RHS is

$$\begin{bmatrix} 2/3 \\ -1 \end{bmatrix}$$

and solving by back substitution gives

$$\mathbf{x} = \begin{bmatrix} -2/9 \\ -1/6 \end{bmatrix}$$

4. (30 points) Let  $A \in \mathbb{C}^{m \times m}$  be a square matrix. Order the singular values  $\sigma_i$  of  $A$  by  $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_m \geq 0$  and order the eigenvalues  $\lambda_i$  of  $A$  so  $|\lambda_1| \geq |\lambda_2| \geq \dots \geq |\lambda_m| \geq 0$ .

(a) Show that  $\sigma_1 \geq |\lambda_1|$ .

Let  $\mathbf{x} \in \mathbb{C}^m$  be the eigenvector associated to  $\lambda_1$  so that  $A\mathbf{x} = \lambda_1\mathbf{x}$ . We may normalize  $\mathbf{x}$  so that  $\|\mathbf{x}\|_2 = 1$ . We then have

$$\|A\mathbf{x}\|_2 = \|\lambda_1\mathbf{x}\|_2 = |\lambda_1|\|\mathbf{x}\|_2 = |\lambda_1|$$

As a consequence,

$$|\lambda_1| \leq \sup_{\mathbf{x} \neq 0} \frac{\|A\mathbf{x}\|_2}{\|\mathbf{x}\|_2} = \|A\|_2 = \sigma_1$$

- (b) Show that  $\sigma_m \leq |\lambda_m|$  (Hint: Write  $\mathbf{x}_m$ , an eigenvector associated to  $\lambda_m$ , in terms of the right singular vectors  $\mathbf{v}_1, \dots, \mathbf{v}_m$  of  $A$ ).

Let  $A = U\Sigma V^*$  be a SVD of  $A$ . We let  $\mathbf{v}_1, \dots, \mathbf{v}_m$  be the columns of  $V$ , i.e. the right singular vectors. Now suppose that  $\mathbf{x}_m$  is a (non-zero) vector so that  $A\mathbf{x}_m = \lambda_m\mathbf{x}_m$ . We may normalize so that  $\|\mathbf{x}_m\|_2 = 1$ . Writing  $\mathbf{x}_m$  in terms of the ONB given by the  $\mathbf{v}_i$  one has

$$\mathbf{x}_m = \sum_{i=1}^m c_i \mathbf{v}_i$$

Note that since  $\|\mathbf{x}_m\|_2 = 1$  one has  $\sum_{i=1}^m |c_i|^2 = 1$  by the Pythagorean theorem. As a consequence  $A\mathbf{x}_m = \sum_{i=1}^m \sigma_i c_i \mathbf{u}_i$ . Hence by the Pythagorean theorem

$$|\lambda_m|^2 = \|A\mathbf{x}_m\|_2^2 = \sum_{i=1}^m \sigma_i^2 |c_i|^2$$

Since  $\sigma_i \geq \sigma_1$  one obtains

$$|\lambda_m|^2 \geq \sum_{i=1}^m \sigma_1^2 |c_i|^2 = \sigma_1^2.$$

- (c) Using part a), show that if  $\|A\|_2 < 1$  then  $I + A$  is nonsingular. Here  $\|A\|_2$  is the induced 2-norm and

$$I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

is the  $3 \times 3$  identity matrix.

(Note that  $I$  should really be the  $m \times m$  identity otherwise the problem makes no sense.) Since  $\|A\|_2 < 1$  we have that  $\sigma_1 < 1$  and so by part a) we have that  $|\lambda_1| < 1$ . In particular, if  $\lambda \in \Lambda(A)$  then  $|\lambda| < 1$ . Since  $\lambda \in \Lambda(I + A)$  if and only if  $\lambda - 1 \in \Lambda(A)$  and  $1 > |\lambda - 1| \geq 1 - |\lambda|$  we see that  $|\lambda| > 0$ . In other words no eigenvalue of  $I + A$  is zero. As a consequence  $N(I + A) = \{0\}$  and so  $I + A$  is invertible.

5. (20 points) Let  $A \in \mathbb{C}^{3 \times 3}$  be the matrix

$$A = \begin{bmatrix} -1 & 3 & -2 \\ 0 & 3 & 1 \\ 0 & 4 & 1 \end{bmatrix}$$

Find a unitary matrix  $Q \in \mathbb{C}^{3 \times 3}$  so that

$$QA = \begin{bmatrix} 1 & * & * \\ 0 & * & * \\ 0 & 0 & * \end{bmatrix}.$$

Here  $*$  represents an unspecified number.

Notice that the first column of  $A$  is  $-\mathbf{e}_1$  and of  $QA$  is  $\mathbf{e}_1$ . Hence, we must have  $Q(-\mathbf{e}_1) = \mathbf{e}_1$ . That is the first column of  $Q$  is  $-\mathbf{e}_1$ . For  $Q$  to be unitary it must have orthonormal columns and hence  $Q$  has the form

$$Q = \begin{bmatrix} -1 & 0 \\ 0 & Q' \end{bmatrix}$$

where  $Q' \in \mathbb{C}^{2 \times 2}$  is unitary. In addition, we want to have

$$Q' \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \begin{bmatrix} x \\ 0 \end{bmatrix}$$

Now for  $Q'$  to be unitary it must preserve length. In particular, one must have  $|x| = \sqrt{3^2 + 4^2} = 5$ . We take  $x = 5$ .

We now have a number of choices we could make in finding  $Q'$ . In the spirit of the Householder algorithm we take  $Q'$  to be a reflection. In this, case we set

$$\mathbf{v} = \begin{bmatrix} 3 \\ 4 \end{bmatrix} - \begin{bmatrix} 5 \\ 0 \end{bmatrix} = \begin{bmatrix} -2 \\ 4 \end{bmatrix}$$

be the vector normal to the line midway between  $[3, 4]^T$  and  $[5, 0]^T$ . The orthonormal projection onto  $\text{span}(\mathbf{v})$  is given by

$$P = \frac{v v^*}{\|v\|_2^2} = \frac{1}{5} \begin{bmatrix} 1 & -2 \\ -2 & 4 \end{bmatrix}$$

Then we have (as we saw in class or can easily convince ourselves) that  $Q'$  is given by

$$Q' = I - 2P = \frac{1}{5} \begin{bmatrix} 3 & 4 \\ 4 & -3 \end{bmatrix}$$

Which we verify has the desired properties. Hence

$$Q = \frac{1}{5} \begin{bmatrix} -5 & 0 & 0 \\ 0 & 3 & 4 \\ 0 & 4 & -3 \end{bmatrix}$$

6. (30 points) (a) Let  $T \in \mathbb{C}^{m \times m}$  be upper triangular. Show that if  $T$  is unitary then  $T$  is diagonal. (Hint: Use the fact that columns are orthogonal and induct on  $m$ ).

We prove the result by induction on  $m$ . When  $m = 1$  then any matrix is diagonal so we are done. Assume the result is true for all  $m \times m$  upper-triangular matrices. We wish to prove it true for  $(m + 1) \times (m + 1)$  upper-triangular matrices.

To that end we note that if  $T \in \mathbb{C}^{(m+1) \times (m+1)}$  is upper triangular and unitary. Then the first column of  $T$  must be of the form  $\mu \mathbf{e}_1$  (as  $T$  upper triangular) and the length of the first column is 1 (as  $T$  unitary) so  $|\mu| = 1$ . Since every other column of  $T$  is orthogonal to the first column (as  $T$  is unitary)  $T$  has the form

$$T = \begin{bmatrix} \mu & 0 \\ 0 & T' \end{bmatrix}.$$

Where  $T' \in \mathbb{C}^{m \times m}$  is upper triangular and unitary. In particular, by the induction hypothesis  $T'$  is diagonal and hence so is  $T$ .

- (b) Let

$$A = \begin{bmatrix} a_{11} & 0 & \cdots & 0 \\ 0 & a_{22} & \cdots & 0 \\ \vdots & & \ddots & \vdots \\ 0 & \cdots & 0 & a_{mm} \end{bmatrix} \in \mathbb{C}^{m \times m}$$

be a diagonal matrix. Show that if  $A$  is unitary then  $|a_{ii}| = 1$  for  $1 \leq i \leq m$ .

Since the columns of a unitary matrix must be of unit length it is straightforward to see that  $|a_{ii}| = 1$ .

- (c) Let  $X \in \mathbb{C}^{m \times m}$  be unitary, use parts a), b) and the Schur factorization to show that  $X$  is unitarily diagonalizable (i.e.  $\mathbb{C}^m$  has an orthonormal basis of eigenvectors) and that  $\lambda \in \Lambda(X)$  implies  $|\lambda| = 1$ .

Write the Schur factorization of  $X$  as

$$X = QTQ^*$$

where  $Q \in \mathbb{C}^{m \times m}$  is unitary and  $T \in \mathbb{C}^{m \times m}$  is diagonal. For  $X$  to be unitary one has  $X^* = X^{-1}$ . On the one hand

$$X^* = (QTQ^*)^* = (Q^*)^* T^* Q^* = QT^* Q.$$

on the other

$$X^{-1} = (QTQ^*)^{-1} = (Q^*)^{-1} T^{-1} Q^{-1} = QT^{-1} Q^*.$$

Hence  $T$  is unitary and so by a) and b) is diagonal with entries on the diagonal all of length 1. In all cases the diagonal entries of  $T$  are the eigenvalues of  $X$  and in this case the columns of  $Q$  are the eigenvectors of  $X$  and so we have proved the claim.

7. (30 points) Let  $A \in \mathbb{R}^{3 \times 3}$  be the matrix

$$A = \begin{bmatrix} -3 & 2 & -2 \\ 0 & 1 & 0 \\ 2 & -1 & 2 \end{bmatrix}$$

(a) Find the eigenvalues of  $A$  and give their algebraic multiplicity.

By expanding along the middle column we see that the characteristic polynomial of  $A$  is:

$$p_A(z) = \det(zI - A) = (z - 1)((z + 3)(z - 2) - 2(-2)) = (z - 1)(z^2 + z - 6 + 4).$$

By inspection (or using the quadratic formula) this can be factored as

$$p_A(z) = (z - 1)(z - 1)(z + 2).$$

Thus, the eigenvalues are  $\lambda = -2$  with algebraic multiplicity 1 and  $\lambda = 1$  with algebraic multiplicity 2.

(b) Verify that  $A$  is diagonalizable and find a basis of eigenvectors.

As

$$A - (-2)I = \begin{bmatrix} -1 & 2 & -2 \\ 0 & 3 & 0 \\ 2 & -1 & 4 \end{bmatrix}$$

we can do Gaussian elimination to see that an eigenvector associated to  $\lambda = -2$  is

$$\mathbf{x}_1 = \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}$$

Similarly, as

$$A - I = \begin{bmatrix} -4 & 2 & -2 \\ 0 & 0 & 0 \\ 2 & -1 & 1 \end{bmatrix}$$

Gaussian elimination shows that one has (linearly independent) eigenvectors

$$\mathbf{x}_2 = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \mathbf{x}_3 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$$

one verifies that  $\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3$  form a linearly independent set and since  $\dim \mathbb{C}^3 = 3$  they must form a basis.

- (c) Determine the matrices  $X$  and  $\Lambda$  so that  $X$  is non-singular and  $\Lambda$  is diagonal and so one has a factorization:

$$A = X\Lambda X^{-1}$$

If we let

$$X = \begin{bmatrix} 1 & 0 & -2 \\ 2 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}.$$

Using Gaussian elimination one computes

$$X^{-1} = \frac{1}{3} \begin{bmatrix} -1 & 2 & -2 \\ 2 & -1 & 4 \\ -2 & 1 & -1 \end{bmatrix}$$

Hence using the fact that the columns of  $X$  are eigenvectors of  $A$  one has

$$A = \frac{1}{3} \begin{bmatrix} 1 & 0 & -2 \\ 2 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -2 \end{bmatrix} \begin{bmatrix} -1 & 2 & -2 \\ 2 & -1 & 4 \\ -2 & 1 & -1 \end{bmatrix}$$

- (d) Compute  $A^n \mathbf{e}_1$  for  $n \geq 1$  an integer. Please simplify your answer as much as possible. Here

$$\mathbf{e}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

As  $A = X\Lambda X^{-1}$  one computes that

$$A^n = X\Lambda^n X^{-1}.$$

That is

$$A^n = \frac{1}{3} \begin{bmatrix} 1 & 0 & -2 \\ 2 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & (-2)^n \end{bmatrix} \begin{bmatrix} -1 & 2 & -2 \\ 2 & -1 & 4 \\ -2 & 1 & -1 \end{bmatrix}$$

Hence,

$$\begin{aligned} A^n \mathbf{e}_1 &= \frac{1}{3} \begin{bmatrix} 1 & 0 & -2 \\ 2 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} -1 \\ 2 \\ (-2)^{n+1} \end{bmatrix} \\ &= \frac{1}{3} \begin{bmatrix} -1 + (-2)^{n+2} \\ 0 \\ 2 + (-2)^{n+1} \end{bmatrix} \end{aligned}$$

8. (30 points) Let  $A \in \mathbb{R}^{2 \times 2}$  be the following matrix

$$A = \begin{bmatrix} 3 & 2 \\ 1 & -2 \end{bmatrix}$$

Let  $S_p = \{\mathbf{x} \in \mathbb{R}^2 : \|\mathbf{x}\|_p = 1\}$ . Let  $AS_p = \{A\mathbf{x} \in \mathbb{R}^2 : \mathbf{x} \in S_p\}$ . Here  $1 \leq p \leq \infty$  and  $\|\cdot\|_p$  is the  $p$ -norm on  $\mathbb{R}^2$ .

- (a) Compute  $\mu_1 = \|A\|_1$  the induced 1-norm of  $A$  and  $\mu_\infty = \|A\|_\infty$  the induced  $\infty$ -norm of  $A$ . Remember to justify your computation.

For a vector  $\mathbf{x} \in S_1$  let us write  $\mathbf{x} = \alpha\mathbf{e}_1 + (1 - \alpha)\mathbf{e}_2$  where we assume  $0 \leq \alpha \leq 1$  (so we assume  $\mathbf{x}$  in first quadrant, by symmetry this is enough). Then  $A\mathbf{x} = (2 + \alpha)\mathbf{e}_1 + (-2 + 3\alpha)\mathbf{e}_2$  then  $\|A\mathbf{x}\|_1 = |2 + \alpha| + |-2 + 3\alpha|$ . When  $\alpha > 2/3$  this is equal to  $2 + \alpha - 2 + 3\alpha = 4\alpha$ , while for  $\alpha \leq 2/3$  this is equal to  $2 + \alpha + 2 - 3\alpha = 4 - 2\alpha$ . Notice that this is maximized for  $\alpha = 0$  or  $\alpha = 1$  and has maximum value  $\mu_1 = 4$ .

For a vector  $\mathbf{y} \in S_\infty$  let us write  $\mathbf{y} = x\mathbf{e}_1 + y\mathbf{e}_2$  where  $x = 1$  and  $0 \leq y \leq 1$  or  $y = 1$  and  $0 \leq x < 1$  (so again we are in the first quadrant). Then  $A\mathbf{y} = (3x + 2y)\mathbf{e}_1 + (x - 2y)\mathbf{e}_2$ . Then  $\|A\mathbf{y}\|_\infty = \max\{|3x + 2y|, |x - 2y|\}$ . By inspection one sees that the maximum value is  $\mu_\infty = 5$ .

- (b) Determine all vectors  $\mathbf{x}_1 \in S_1$  and  $\mathbf{x}_\infty \in S_\infty$  so that  $\|A\mathbf{x}_1\|_1 = \mu_1$  and  $\|A\mathbf{x}_\infty\|_\infty = \mu_\infty$ .

In the previous problem we see that  $\mathbf{x}_1$  can be any of the following and no other vectors:

$$\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ -1 \end{bmatrix}$$

In the previous problem we see that  $\mathbf{x}_\infty$  can be any of the following and no other vectors:

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ -1 \end{bmatrix},$$

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(c) Sketch  $S_1$  and  $AS_1$  and indicate the vectors  $\mathbf{x}_1$  and  $A\mathbf{x}_1$  on your picture.

(d) Sketch  $S_\infty$  and  $AS_\infty$  and indicate the vectors  $\mathbf{x}_\infty$  and  $A\mathbf{x}_\infty$  on your picture.