

Revised 10/23/03

LAB 5: Hermitian and Normal Matrices, Positive-Definite Matrices, and the Singular Value Decomposition

In this lab you will use MATLAB to study four topics:

- The eigenvectors/eigenvalues of hermitian and normal matrices
- Tests for positive-definiteness of a real symmetric matrix
- The Cholesky Decomposition of a Positive-definite matrix
- The singular value decomposition of a matrix, which expresses the action of a matrix as a rotation, followed by a stretching in each coordinate direction, followed by another rotation.

Tcodes: For this lab you will need the Teaching Codes

```
house.m, plot2d.m, slu.m
```

Before opening MATLAB to work on the Lab questions you should copy these codes to your directory. See Lab #2 for further details on how to obtain any of these files from the course home page.

Random Seed: When you start your MATLAB session, initialize the random number generator by typing

```
rand('seed', abcd)
```

where $abcd$ are the last four digits of your Student ID number. This will ensure that you generate your own particular random vectors and matrices.

BE SURE TO INCLUDE THIS LINE IN YOUR LAB WRITE-UP

Question 1. Diagonalization of Normal Matrices

For this question first generate random 4×4 complex matrices A and B by

```
A =rand(4) + i*rand(4); B = (A + A')/2
```

(here $i = \sqrt{-1}$). Note that if A is a complex matrix, then the MATLAB notation A' means the *conjugate* transpose A^H . Hence B is hermitian symmetric.

(a) Compute the matrix S of eigenvectors of A and the diagonal eigenvalue matrix L of A by

```
[S, L] = eig(A)
```

Since A is a random matrix, it should have 4 distinct eigenvalues. Verify by MATLAB that $S^{-1}AS = L$. Are the eigenvectors of A mutually orthogonal? Is $S' * A * S = L$? Is $A * A' = A' * A$? Explain the relations among these questions.

(b) Compute the matrix U of eigenvectors of B and the diagonal eigenvalue matrix of B by

```
[U, D] = eig(B)
```

What can you predict about the eigenvalues of B without calculation? Are the eigenvectors of B mutually orthogonal? Confirm your predictions with MATLAB calculations and verify that $U^* B U = D$.

(c) Generate a random diagonal matrix

$$F = \text{diag}(\text{diag}(\text{rand}(4) + i * \text{rand}(4))).$$

Change F into a non-diagonal matrix by setting $C = U * F * U^*$ (where U is the matrix from part (b)). Prove (by matrix algebra without numerical calculation) that $BC = CB$ and $C^H C = C C^H$. Then use MATLAB to check these equations.

(d) Calculate the eigenvalues and eigenvectors of C using

$$[W, E] = \text{eig}(C)$$

What is the relation between E and F ? Are the eigenvectors of C mutually orthogonal? Are the columns of W eigenvectors for B ? Answer these last two questions by general theory, and then confirm by MATLAB calculations.

Question 2. Positive-Definite Matrices

(a) **Tests for Positivity:** (see Strang, page 331, 6B) Generate a random 3×3 real symmetric matrix A by

$$B = \text{fix}(5 * \text{rand}(3)); A = B + B'$$

Determinant Test: Calculate the three *upper left determinants* of A by

$$D_1 = A(1,1), D_2 = \det(A(1:2,1:2)), D_3 = \det(A)$$

One test for A to be *positive definite* is

$$D_1 > 0, \quad D_2 > 0, \quad D_3 > 0$$

Does the matrix A pass this test?

Eigenvalue Test: Another test for A to be positive definite is that all the *eigenvalues* of A are positive. Calculate these eigenvalues by the MATLAB command `eig(A)`. Are the results of this test consistent with the Determinant Test?

Pivot Test: A third test for A to be positive definite is that all the *pivots* in the *LU* decomposition of A are positive. Check this by calculating

$$[L, U] = \text{slu}(A)$$

The pivots are the diagonal entries of U (here `slu.m` is the Teaching Code that you used in Lab 2). Are the pivots all positive? Is this consistent with the previous two tests?

If the matrix A you generated failed any of these tests, generate another symmetric matrix until you get one that passes all the tests.

(b) **Positivity of Quadratic Form $x^T A x$:** Let A be a matrix from (a) that passed all the tests for positivity. Generate several random vectors $x = \text{rvect}(3)$ and check that $x^* A x$ is positive for each one.

(c) **Law of Inertia:** Generate a random 3×3 real symmetric matrix A by the method of part (a) that is *not* positive definite (you can use one of the matrices that failed the tests in (a)).

Verify by MATLAB that A has the same number of negative pivots as negative eigenvalues (See Strang, page 342, 6G). (In counting the number of negative pivots, remember that the pivots are $D_1, D_2/D_1, D_3/D_2$.)

Question 3. Cholesky Factorization of Positive-Definite Matrices

(a) If B has independent columns, then the matrix $A = B^T B$ is always positive definite. Check this by generating a random 4×3 matrix $B = \text{rmat}(4,3)$. Calculate `rank(B)` to determine if B has independent columns. If this is the case, set $A = B' * B$ (otherwise, generate another matrix B and repeat). Next, calculate `eig(A)` to see that A is positive definite. Generate several random vectors $\mathbf{x} = \text{rvect}(3)$ and check that $\mathbf{x}' * A * \mathbf{x}$ is positive for each one.

(b) Use MATLAB to calculate an orthonormal basis of eigenvectors for A by the command

$$[Q, D] = \text{eig}(A).$$

Use MATLAB to verify that Q is an orthogonal matrix and that $Q' * A * Q = D$.

(c) The positive-definite matrix A has a *Cholesky factorization* $A = R^T R$, where R is upper triangular with positive diagonal entries. One way to obtain this factorization is to start with the *LU* factorization. Compute

$$[L, U] = \text{slu}(A), \quad H = \text{diag}(\text{diag}(U)) .$$

Show by hand calculation why $A = LHL^T$, and then verify this by MATLAB. The matrices A and H are congruent. Do they have the same eigenvalues? Explain. Now compute $R = \text{sqrt}(H) * L'$. Show by hand calculation why $A = R^T * R$. Then verify this by MATLAB. Finally, calculate the Cholesky decomposition of A by the the MATLAB m-file `chol(A)` and verify that you get the same matrix R . (The built-in MATLAB function takes advantage of the positive-definiteness of A and is considerably faster than the *LU* factorization method.)

Question 4. Singular Value Decomposition

(a) **Graphic Demo of SVD:** Generate a random 2×2 matrix $A = \text{rand}(2,2)$. Then type `eigshow(A)` at the MATLAB prompt. A graphics window should open. Click on the `svd` button on the right side of the window. Your matrix A should appear (in MATLAB notation) in the menu bar above the graph. Underneath the graph the statement

Make $A * \mathbf{x}$ perpendicular to $A * \mathbf{y}$

should appear (if it does not, then click on the `svd` button again). The graph shows a pair of orthogonal unit vectors \mathbf{x} and \mathbf{y} , together with the transformed vectors $A\mathbf{x}$ and $A\mathbf{y}$.

Move the pointer onto the vector \mathbf{x} , and then make the pair of vectors \mathbf{x} , \mathbf{y} go around a circle. The transformed vectors $A\mathbf{x}$ and $A\mathbf{y}$ then move around an ellipse, but generally $A\mathbf{x}$ is not perpendicular to $A\mathbf{y}$. Search for the position of \mathbf{x} and \mathbf{y} so that $A\mathbf{x}$ is perpendicular to $A\mathbf{y}$. When this happens, then the *singular values* σ_1 and σ_2 of A are the lengths of the vectors $A\mathbf{x}$ and $A\mathbf{y}$. Give a rough estimate of these lengths from the graph (remember that \mathbf{x} and \mathbf{y} have length one).

(b) **Calculation of SVD:** Let A be the random matrix you generated in part (a). Use MATLAB to calculate the singular value decomposition of A by

$$[U, S, V] = \text{svd}(A)$$

Verify that $A = U * S * V'$ (up to numerical roundoff error).

The diagonal entries of S are the *singular values* σ_1, σ_2 of A . Compare the calculated singular values with your graphical estimates for σ_1, σ_2 from part (a).

The singular values are the square roots of the eigenvalues of $A^T A$. Check this by calculating

```
sqrt(eig(A'*A))
```

(c) **Geometric Meaning of SVD:** At the MATLAB prompt type

```
H = house; plot2d(H)
```

A graphics window should open and display a crude drawing of a house (the matrix H contains the coordinates of the line segments of the figure).

Generate an orthogonal matrix V by

```
t = pi/6; V = [cos(t), -sin(t); sin(t), cos(t)]
```

Let V^T act on the house by `plot2d(V'*H)` (be sure to use the transpose matrix V'). How has the house been changed?

Next generate a diagonal matrix S by

```
S = [5/4, 0; 0, 3/4 ]
```

Let S act on the rotated house by `plot2d(S*V'*H)`. How does this change the previous picture?

Now generate another orthogonal matrix U by

```
t = pi/4; U = [cos(t), -sin(t); sin(t), cos(t)]
```

Let U act on the rotated and stretched house by `plot2d(U*S*V'*H)`. How does this change the previous picture? Print this figure and include it in your lab write up.

Set $A = U*S*V'$. Use MATLAB to calculate the singular value decomposition of A by

```
[U1, S1, V1] = svd(A)
```

Verify that $S = S1$ and that $U1*S1*V1' = A$ (up to numerical roundoff error). Note that the U, V matrices in the svd are *not* unique.