

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 disk space. See Lab #2 for further details on how to obtain any of these files from the course home page.

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 = diag(diag(rand(4)+i*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 $x' * R * x$ 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] = 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] = slu(A), H = diag((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 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*x perpendicular to A*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] = 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 $U = U1$, $S = S1$, $V = V1$ (up to numerical roundoff error).