Problem 1

Show that every positive definite matrix must be invertible. Also show that if a matrix is positive definite, then it is unitary if and only if all of it's eigenvalues are 1.

Solution

Proof. Let A be a positive definite $n \times n$. By definition, this matrix is hermitian, and therefore there exists an orthonormal basis $\beta = \{v_1, \dots, v_n\}$ of our vector space consisting of eigenvectors of A, with corresponding eigenvalues λ_i . By problem 2.a of HW2, we have that $\lambda_i > 0$ for all i. Let $x \in \mathbb{F}^n$ be nonzero. Then, there exist $a_i \in \mathbb{F}$, at least one of which is nonzero, such that

$$x = \sum_{i=1}^{n} a_i v_i.$$

Then

$$Ax = \sum_{i=1}^{n} a_i A v_i$$
$$= \sum_{i=1}^{n} a_i \lambda_i v_i.$$

Since each λ_i and at least one a_i is nonzero, the linear independence of β tells us that Ax is nonzero. Thus, the dimension of the nullspace of A is zero, and we can conclude that A is invertible.

Now, suppose A is also unitary. Then, we have

$$\langle Av_i, Av_i \rangle = \lambda_i^2$$

= $\langle x, x \rangle$ Property of unitary operators = 1 Since our basis is orthonormal.

Thus, since λ_i is real and nonnegative, we can conclude $\lambda_i = 1$ for each i.

For the other direction, suppose As eigenvalues are all 1. Then, defining x as we did above, we have

$$\langle Ax, Ax \rangle = \sum_{i=1}^{n} |a_i|^2 \lambda_i^2$$
$$= \sum_{i=1}^{n} |a_i|$$
$$= \langle x, x \rangle,$$

and we have shown that A is unitary.

Problem 2

Use question 1 and the previous paragraph to solve the following generalized eigenvalue problem $Ax = \lambda Bx$.

$$\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} x = \lambda \begin{bmatrix} 3 & -1 \\ -1 & 1 \end{bmatrix} x$$

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Solution

Since B has a nonzero determinant, we can see it is invertible. Thus, we will find it's inverse:

$$[B \quad I] = \begin{bmatrix} 3 & -1 & 1 & 0 \\ -1 & 1 & 0 & 1 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} 1 & 1 & 1 & 2 \\ -1 & 1 & 0 & 1 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} 1 & 1 & 1 & 2 \\ 0 & 2 & 1 & 3 \end{bmatrix}$$

$$R_1 + 2R_2$$

$$R_2 + R_1$$

$$\rightarrow \begin{bmatrix} 1 & 1 & 1 & 2 \\ 0 & 1 & \frac{1}{2} & \frac{3}{2} \end{bmatrix}$$

$$\frac{1}{2}R_2$$

$$\rightarrow \begin{bmatrix} 1 & 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & 1 & \frac{1}{2} & \frac{3}{2} \end{bmatrix}$$

$$R_1 - R_2$$

Thus, we have

$$B^{-1} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{3}{2} \end{bmatrix}.$$

Now we compute $B^{-1}A$:

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

Our generalized eigenvalue problem reduces to finding the eigenvectors of this matrix:

$$|B^{-1}A - \lambda I| = \begin{vmatrix} \frac{1}{2} - \lambda & 1\\ \frac{1}{2} & 2 - \lambda \end{vmatrix}$$
$$= (\frac{1}{2} - \lambda)(2 - \lambda) - \frac{1}{2}$$
$$= \lambda^2 - \frac{5}{2}\lambda + \frac{1}{2}.$$

From the quadratic formula, we have eigenvalues

$$\lambda_1 = \frac{1}{4}(5 - \sqrt{17})$$
, and $\lambda_2 = \frac{1}{4}(5 + \sqrt{17})$.

Now we find our eigenvectors:

$$\begin{bmatrix} \frac{1}{2} - \lambda_1 & 1 \\ \frac{1}{2} & 2 - \lambda_1 \end{bmatrix} = \begin{bmatrix} \frac{1}{2} - \frac{1}{4}(5 - \sqrt{17}) & 1 \\ \frac{1}{2} & 2 - \frac{1}{4}(5 - \sqrt{17}) \end{bmatrix}$$

$$= \begin{bmatrix} \frac{-3}{4} + \frac{1}{4}\sqrt{17} & 1 \\ \frac{1}{2} & 2 - \frac{1}{4}(5 - \sqrt{17}) \end{bmatrix} .$$

Thus, looking at the first row, we have an eigenvector

$$v_1 = \begin{bmatrix} 4\\ 3 - \sqrt{17} \end{bmatrix}.$$

For the next one, we follow the same procedure:

$$\begin{bmatrix} \frac{1}{2} - \lambda_2 & 1 \\ \frac{1}{2} & 2 - \lambda_2 \end{bmatrix} = \begin{bmatrix} \frac{1}{2} - \frac{1}{4}(5 + \sqrt{17}) & 1 \\ \frac{1}{2} & 2 - \frac{1}{4}(5 + \sqrt{17}) \end{bmatrix}$$

$$= \begin{bmatrix} \frac{-3}{4} - \frac{1}{4}\sqrt{17} & 1 \\ \frac{1}{2} & 2 - \frac{1}{4}(5 - \sqrt{17}) \end{bmatrix} .$$

Again, the first row gives us the eigenvector

$$v_2 = \begin{bmatrix} 4\\ 3 + \sqrt{17} \end{bmatrix},$$

and we have solved the generalized eigenvalue problem.

Problem 3

Solve the generalized eigenvalue problem from Question 2 using this second method.

Solution

We have

$$|A - \lambda B| = \begin{vmatrix} 1 & 1 \\ 0 & 1 \end{vmatrix} - \lambda \begin{bmatrix} 3 & -1 \\ -1 & 1 \end{bmatrix}$$

$$= \begin{vmatrix} 1 & 1 \\ 0 & 1 \end{vmatrix} - \begin{bmatrix} 3\lambda & -\lambda \\ -\lambda & \lambda \end{bmatrix} \begin{vmatrix} 1 & 1 \\ -\lambda & \lambda \end{vmatrix}$$

$$= \begin{vmatrix} 1 - 3\lambda & 1 + \lambda \\ \lambda & 1 - \lambda \end{vmatrix}$$

$$= (1 - 3\lambda)(1 - \lambda) - \lambda(1 + \lambda)$$

$$= 3\lambda^2 - 4\lambda + 1 - \lambda - \lambda^2$$

$$= 2\lambda^2 - 5\lambda + 1$$

$$= 0$$

$$0 = \lambda^2 - \frac{5}{2}\lambda + \frac{1}{2}.$$

As before, this is the same equation we had before (as it clearly should be), so we have eigenvalues

$$\lambda_1 = \frac{1}{4}(5 - \sqrt{17}), \text{ and } \lambda_2 = \frac{1}{4}(5 + \sqrt{17}).$$

Now we find the eigenvectors. First, we do a little reduction

$$\begin{bmatrix}
1 - 3\lambda & 1 + \lambda \\
\lambda & 1 - \lambda
\end{bmatrix} \rightarrow \begin{bmatrix}
1 - 2\lambda & 2 \\
\lambda & 1 - \lambda
\end{bmatrix} \qquad R_1 + R_2$$

$$\rightarrow \begin{bmatrix}
\frac{1}{2} - \lambda & 1 \\
\lambda & 1 - \lambda
\end{bmatrix} \qquad \frac{1}{2}R_1$$

$$\rightarrow \begin{bmatrix}
\frac{1}{2} - \lambda & 1 \\
\frac{1}{2} & 2 - \lambda
\end{bmatrix} \qquad R_2 + R_1.$$

We reduced this to the same matrix we had set equal to $B^{-1}A - \lambda I$ in problem 2. Since we have already shown the eigenvalues to be the same, we can conclude that the eigenvectors are the same (as they should be).

Problem 4

For two $n \times n$ complex matrices A and B, what is the maximum number of distinct generalized eigenvalues we would expect? Justify your answer. Then solve the generalized eigenvalue problem corresponding with the matrix pencil

$$\left(\begin{bmatrix}0&1\\0&0\end{bmatrix},\begin{bmatrix}1&0\\0&0\end{bmatrix}\right).$$

Solution

Theorem 1. For two $n \times n$ complex matrices A and B, the maximum number of distinct generalized eigenvalues is n.

Proof. We claim that the determinant of $A - \lambda B$ is at most an n^{th} degree polynomial of λ , from which the conclusion follows. We will proceed by induction on n. For n = 2, this is immediately clear.

Now, suppose for some $n \geq 2$ that for any two $n \times n$ complex matrices, the maximum number of distinct generalized eigenvalues is n. Let A and B be $n+1\times n+1$ complex matrices. For an $m\times m$ matrix C, let \overline{C}_{ij} denote the $m-1\times m-1$ submatrix of C obtained by deleting the i^{th} row and j^{th} column of C. Then, we have

$$|A - \lambda B| = \sum_{j=1}^{n+1} (A_{1j} - \lambda B_{1j}) |\overline{A - \lambda B}_{1j}|.$$

By the inductive hypothesis, $|\overline{A-\lambda B_{1j}}|$ is an at most $n \times n$ polynomial of λ . Since $(A_{1j} - \lambda B_{1j})$ is an at most first degree polynomial, we have that each term of the above sum is an at most $n+1^{\text{th}}$ degree polynomial, and we have completed our proof.

Now, we solve the generalized eigenvalue problem:

$$|A - \lambda B| = \begin{vmatrix} -\lambda & 1\\ 0 & 0 \end{vmatrix}$$
$$= 0$$

Thus, every nonzero vector is a generalized eigenvector with corresponding eigenvalue 0.

Problem 5

Theorem 2. Let \hat{E}_{λ} be the collection of all generalized eigenvectors associated with the eigenvalue λ . Then \hat{E}_{λ} is a subspace of V.

Proof. We have

$$\hat{E}_{\lambda} = \{ x \in V | (T - \lambda U)x = 0 \}.$$

That is, \hat{E}_{λ} is the nullspace of the linear operator $(T - \lambda U)$, and is therefore a subspace of V.

Problem 6

For distinct generalized eigenvalues λ and ν it is not necessarily the case that $\hat{E}_{\lambda} \cap \hat{E}_{\nu} = \{0\}$. Given a regular matrix pencil (A, B) and distinct nonzero eigenvalues $\lambda \neq \nu$, what is $\hat{E}_{\lambda} \cap \hat{E}_{\nu}$?

Solution

Let $x \in \hat{E}_{\lambda} \cap \hat{E}_{\nu}$. Then, we have

$$(A - \lambda B)x = 0$$

$$= (A - \nu B)x$$

$$0 = (A - \lambda B)x - (A - \nu B)x$$

$$= Ax - \lambda Bx - Ax + \nu Bx$$

$$= (\nu - \lambda)Bx,$$

and we can conclude that $x \in N(B)$. With this, we have

$$0 = (A - \lambda B)x$$
$$= Ax - \lambda Bx$$
$$= Ax.$$

Thus, we also have that $x \in N(A)$. Therefore, we have $x \in N(A) \cap N(B)$ which implies $\hat{E}_{\lambda} \cap \hat{E}_{\nu} \subseteq N(A) \cap N(B)$. Supposing that $x \in N(A) \cap N(B)$, it follows immediately that $x \in \hat{E}_{\lambda} \cap \hat{E}_{\nu}$. With this, we can conclude that $\hat{E}_{\lambda} \cap \hat{E}_{\nu} = N(A) \cap N(B)$.

Problem 7

Let V be an n-dimensional vector space over \mathbb{C} . Suppose that P and Q are two orthogonal projections. Suppose that L = R(P), and N = R(Q). Then we may write L as

$$L = (L \cap N) \oplus (L \cap N^{\perp}).$$

There is an equivalent decomposition for L^{\perp} . With these two decompositions we can write

$$V = (L \cap N) \oplus (L \cap N^{\perp}) \oplus (L^{\perp} \cap N) \oplus (L^{\perp} \cap N^{\perp})$$

Solution

Proof. By theorem 6.6 in our textbook, we have

$$V = N \oplus N^{\perp}$$
.

since the range of Q is a subspace of V. Thus, we have

$$x \in L \cap V \iff x \in L \cap (N + N^{\perp})$$

$$\iff x \in L \wedge x \in \{x_1 + x_2 | x_1 \in N \wedge x_2 \in N^{\perp}\}$$

$$\iff x \in \{x_1 + x_2 | x_1 \in L \cap N \wedge x_2 \in L \cap N^{\perp}\}$$

$$\iff x \in (L \cap N) + L \cap N^{\perp}.$$

Furthermore, we have

$$x \in (L \cap N) \cap (L \cap N^{\perp}) \iff x \in L \cap (N \cap N^{\perp})$$

 $\iff x \in L \cap \{0\}$ Since Q is an orthogonal projection
 $\iff x = 0$.

Thus, we have shown that

$$L = (L \cap N) \oplus (L \cap N^{\perp}).$$

In a nearly identical argument, we can derive the equality

$$L^{\perp} = (L^{\perp} \cap N) \oplus (L^{\perp} \cap N^{\perp}).$$

Since P is an orthogonal projection, we have

$$V = L \oplus L^{\perp}$$

= $(L \cap N) \oplus (L \cap N^{\perp}) \oplus (L^{\perp} \cap N) \oplus (L^{\perp} \cap N^{\perp}),$

as desired.

Problem 8

What happens to the decomposition in the first part of question 7 if we only assume that Q is a projection, but not an orthogonal projection? Give a concrete example of such a projection onto a subspace of \mathbb{C}^2 .

Note for Dr. Bossaller: you wrote P for this problem instead of Q, but I am pretty certain you meant Q.

Solution

By theorem 6.6 in our textbook, we have $V = W \oplus W^{\perp}$ for any subspace of W of V. If Q was not an orthogonal projection, then we would still be able to write

$$V = N \oplus N^{\perp}$$
.

as N would still be a subspace of V. In fact, this would be true if Q were any operator on V, since the range of any operator is a subspace of V. The notable unique quality of orthogonal projections is that $R(Q)^{\perp} = N(Q)$, but the decomposition above does not explicitly use the null space of Q, thus the decomposition holds.

To see a concrete example, let $V = \mathbb{C}^2$, and consider the projection Q onto Span(1,1) given by

$$Q(x_1, x_2) = (x_1, x_1).$$

Then, we have

$$\begin{split} \langle Qx,y\rangle &= \langle (x_1,x_1),(y_1,y_2)\rangle \\ &= x_1\overline{y_1} + x_1\overline{y_2} \\ &= x_1(\overline{y_1} + \overline{y_2}) \\ &= x_1\overline{(y_1+y_2)}. \end{split}$$

From this, we can see that $N^{\perp}=\{(y,-y)|y\in\mathbb{C}\}=\mathrm{Span}(1,-1).$ Furthermore, we have

$$\begin{split} N(Q) &= \mathrm{Span}(0,1) \\ &\neq N^\perp, \end{split}$$

which means the projection we have chosen is not orthogonal. Finally, we can see that (1,-1) and (0,1) are linearly independent, and if follows that

$$V = N \oplus N^{\perp},$$

as we expect.