

# Abstract Linear Algebra: Homework 6

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**Problem 1.** Let  $T$  be the linear transformation on  $\mathbb{R}^4$  which is represented in standard ordered basis by the matrix

$$\begin{pmatrix} 0 & 0 & 0 & 0 \\ a & 0 & 0 & 0 \\ 0 & b & 0 & 0 \\ 0 & 0 & c & 0 \end{pmatrix}.$$

Under what condition on  $a$ ,  $b$ , and  $c$  is  $T$  diagonalizable? Explain your answer.

*Solution.* Computing the determinant using cofactor expansion along the first column gives us

$$\det(A - \lambda I) = -\lambda \begin{vmatrix} -\lambda & 0 & 0 \\ b & -\lambda & 0 \\ 0 & c & -\lambda \end{vmatrix} = -\lambda [(-\lambda)(\lambda^2) - 0 - 0] = \lambda^4.$$

The roots of the characteristic polynomial are all  $\lambda = 0$  with multiplicity of 4.

To determine diagonalizability, we check the geometric multiplicity of  $\lambda = 0$ , which is the dimension of the null space. This gives the system of equations

$$ax_1 = 0, \quad bx_2 = 0, \quad \text{and} \quad cx_3 = 0.$$

For  $T$  to be diagonalizable, we must have  $\dim(\text{Null}(A)) = 4$ , which happens if and only if  $a = b = c = 0$ . Otherwise, the geometric multiplicity is strictly less than 4, and the matrix is not diagonalizable.  $\square$

**Problem 2.** Let  $T$  be a linear transformation on the  $n$ -dimensional vector space  $V$ , and suppose that  $T$  has  $n$  distinct eigenvalues. Prove that  $T$  is diagonalizable.

*Solution.* Since  $T$  has  $n$  distinct eigenvalues, say  $\lambda_1, \lambda_2, \dots, \lambda_n$ , there exist corresponding eigenvectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  such that for each  $i$ , we get  $T(\mathbf{v}_i) = \lambda_i \mathbf{v}_i$ .

For  $n = 1$ , we get the set  $\{\mathbf{v}_1\}$ , which is a single nonzero vector. Therefore, it is linearly independent.

Assume that for any set of  $n$  distinct eigenvalues  $\lambda_1, \dots, \lambda_n$  with corresponding eigenvectors  $\mathbf{v}_1, \dots, \mathbf{v}_n$ , the set  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is linearly independent. Now consider a set of  $n + 1$  distinct eigenvalues  $\lambda_1, \dots, \lambda_n, \lambda_{n+1}$  with corresponding eigenvectors  $\mathbf{v}_1, \dots, \mathbf{v}_n, \mathbf{v}_{n+1}$ . Suppose there exists a linear dependence among these eigenvectors

$$c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_n \mathbf{v}_n + c_{n+1} \mathbf{v}_{n+1} = \mathbf{0} \tag{1}$$

We need to show that all  $i$ ,  $c_i = 0$ .

Applying  $T$  to both sides and subtracting  $\lambda_{n+1}$  times equation 1 from the left side gives us

$$(c_1 \lambda_1 \mathbf{v}_1 + \dots + c_n \lambda_n \mathbf{v}_n + c_{n+1} \lambda_{n+1} \mathbf{v}_{n+1}) - \lambda_{n+1} (c_1 \mathbf{v}_1 + \dots + c_n \mathbf{v}_n + c_{n+1} \mathbf{v}_{n+1}) = \mathbf{0}.$$

Simplifying gives us

$$c_1 (\lambda_1 - \lambda_{n+1}) \mathbf{v}_1 + \dots + c_n (\lambda_n - \lambda_{n+1}) \mathbf{v}_n = \mathbf{0}.$$

By the inductive hypothesis,  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is linearly independent, so the coefficients must be zero

$$c_1 (\lambda_1 - \lambda_{n+1}) = 0, \quad \dots, \quad c_n (\lambda_n - \lambda_{n+1}) = 0.$$

Since the eigenvalues are distinct,  $\lambda_i - \lambda_{n+1} \neq 0$  for all  $i$ , that implies that  $c_1 = c_2 = \dots = c_n = 0$ . Returning back to equation 1, we see that

$$c_n \mathbf{v}_n + c_{n+1} \mathbf{v}_{n+1} = \mathbf{0}.$$

Since  $\mathbf{v}_{n+1} \neq \mathbf{0}$ , we must have  $c_n = c_{n+1} = 0$ . Therefore, set  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is linearly independent.

Since  $V$  is  $n$ -dimensional and we have found  $n$  linearly independent eigenvectors, they form a basis for  $V$ . With respect to this basis of eigenvectors, the matrix representation of  $T$  is diagonal, with the eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_n$  as the diagonal entries. That is, in this basis,  $T$  is represented by the diagonal matrix

$$D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}.$$

Therefore,  $T$  is diagonalizable.  $\square$

**Problem 3.** Let  $T$  be an invertible linear transformation on a vector space  $V$ . Prove that a scalar  $\lambda$  is an eigenvalue of  $T$  if and only if  $\lambda^{-1}$  is an eigenvalue of  $T^{-1}$ .

*Solution.* By definition,  $\lambda$  is an eigenvalue of  $T$  if there exists a nonzero vector  $\mathbf{v} \in V$  such that  $T(\mathbf{v}) = \lambda\mathbf{v}$ . Since  $T$  is invertible, we can apply  $T^{-1}$  to both sides to get  $\mathbf{v} = \lambda T^{-1}(\mathbf{v})$ . Rearranging gives us  $T^{-1}(\mathbf{v}) = \lambda^{-1}\mathbf{v}$ . Since  $\mathbf{v} \neq \mathbf{0}$ ,  $\lambda^{-1}$  is an eigenvalue of  $T^{-1}$  with eigenvector  $\mathbf{v}$ .

Assume that  $\lambda^{-1}$  is an eigenvalue of  $T^{-1}$ . Then, by definition, there exists a nonzero vector  $\mathbf{v} \in V$  such that  $T^{-1}(\mathbf{v}) = \lambda^{-1}\mathbf{v}$ . Applying  $T$  to both sides, we get  $T(T^{-1}(\mathbf{v})) = T(\lambda^{-1}\mathbf{v}) \Rightarrow \mathbf{v} = \lambda^{-1}T(\mathbf{v})$ . Multiplying both sides by  $\lambda$ , we obtain  $\lambda\mathbf{v} = T(\mathbf{v})$ . Since  $\mathbf{v} \neq \mathbf{0}$ , this shows that  $\lambda$  is an eigenvalue of  $T$  with eigenvector  $\mathbf{v}$ .

Therefore,  $\lambda$  is an eigenvalue of  $T$  if and only if  $\lambda^{-1}$  is an eigenvalue of  $T^{-1}$ .  $\square$

**Problem 4.** Let  $A, B \in \mathbb{R}^{n \times n}$ . This problem is to conclude that  $AB$  and  $BA$  have exactly the same set of eigenvalues.

(i) Assume that  $\lambda I - AB$  is invertible. Prove that

$$(\lambda I - BA) [I + B(\lambda I - AB)^{-1}A] = \lambda I.$$

(ii) Use part (i) to prove that  $AB$  and  $BA$  have the same eigenvalues. (Note: The algebraic multiplicity of the same eigenvalues may not be the same.)

*Solution to (i).* Expanding the left-hand side gives us

$$(\lambda I - BA)I + (\lambda I - BA)B(\lambda I - AB)^{-1}A = \lambda I - BA + (\lambda I - BA)B(\lambda I - AB)^{-1}A.$$

Expanding the second term

$$(\lambda I - BA)B(\lambda I - AB)^{-1}A = \lambda B(\lambda I - AB)^{-1}A - BAB(\lambda I - AB)^{-1}A.$$

Notice that we can rewrite  $BAB(\lambda I - AB)^{-1}A$  as  $B(AB)(\lambda I - AB)^{-1}A$  and since we know that  $AB(\lambda I - AB)^{-1} = I - \lambda(\lambda I - AB)^{-1}$ , we can substitute this back into our original equation to get

$$\lambda I - BA + \lambda B(\lambda I - AB)^{-1}A - B(I - \lambda(\lambda I - AB)^{-1})A.$$

Distributing  $B$  in the last term gives us

$$B(I - \lambda(\lambda I - AB)^{-1})A = BA - B\lambda(\lambda I - AB)^{-1}A.$$

Substituting this back, we get

$$\lambda I - BA + \lambda B(\lambda I - AB)^{-1}A - BA + B\lambda(\lambda I - AB)^{-1}A.$$

Now, observe that the terms  $\lambda B(\lambda I - AB)^{-1}A$  and  $B\lambda(\lambda I - AB)^{-1}A$  are the same, so they cancel out, leaving us with

$$\lambda I - BA + BA - BA = \lambda I.$$

Therefore, we have shown that

$$(\lambda I - BA) [I + B(\lambda I - AB)^{-1}A] = \lambda I. \quad \square$$

*Solution to (ii).* Taking determinants on both sides from the equation proved in part (i), we get

$$\det((\lambda I - BA) [I + B(\lambda I - AB)^{-1}A]) = \det(\lambda I).$$

Using the determinant property  $\det(AB) = \det(A)\det(B)$ , we rewrite this as

$$\det(\lambda I - BA) \cdot \det(I + B(\lambda I - AB)^{-1}A) = \det(\lambda I).$$

If  $\lambda I - AB$  is invertible, we analyze  $I + B(\lambda I - AB)^{-1}A$ . Suppose  $I + B(\lambda I - AB)^{-1}A$  is invertible, then

$$\det(\lambda I - BA) = \frac{\det(\lambda I)}{\det(I + B(\lambda I - AB)^{-1}A)}.$$

Since the right-hand side is nonzero when  $I + B(\lambda I - AB)^{-1}A$  is invertible, this means that  $\lambda I - BA$  is invertible whenever  $\lambda I - AB$  is invertible.

Eigenvalues correspond to values of  $\lambda$  for which the matrix  $\lambda I - AB$  is not invertible, meaning

$$\det(\lambda I - AB) = 0.$$

From our determinant equation, if  $\lambda I - AB$  is singular (i.e.,  $\det(\lambda I - AB) = 0$ ), then  $\lambda I - BA$  must also be singular (i.e.,  $\det(\lambda I - BA) = 0$ ). This shows that if  $\lambda$  is an eigenvalue of  $AB$ , then it is also an eigenvalue of  $BA$ .

Conversely, if  $\lambda I - BA$  is singular, then  $I + B(\lambda I - AB)^{-1}A$  must be non-invertible, implying that  $\lambda I - AB$  is also singular. This proves that if  $\lambda$  is an eigenvalue of  $BA$ , then it is also an eigenvalue of  $AB$ .

Thus,  $AB$  and  $BA$  have exactly the same eigenvalues.  $\square$

**Problem 5.** Let  $A \in \mathbb{C}^{n \times n}$ . Let  $g$  be a polynomial over  $\mathbb{C}$ . Prove that  $c$  is an eigenvalue of  $g(A)$  if and only if  $c = g(\lambda)$  for some eigenvalue  $\lambda$  of  $A$ .

*Solution.* Let  $A \in \mathbb{C}^{n \times n}$  and let  $g$  be a polynomial over  $\mathbb{C}$ . Suppose  $g(A)$  is formed by evaluating  $g$  at  $A$ , meaning that

$$g(A) = \sum_{k=0}^m a_k A^k.$$

Suppose  $c$  is an eigenvalue of  $g(A)$ . By definition, there exists a nonzero vector  $\mathbf{v} \in \mathbb{C}^n$  such that  $g(A)\mathbf{v} = c\mathbf{v}$ . We need to show that  $c = g(\lambda)$  for some eigenvalue  $\lambda$  of  $A$ . To do so, we note that  $\mathbf{v}$  is an eigenvector of  $g(A)$ , corresponding to the eigenvalue  $\lambda$ . To justify this, recall that any polynomial of a matrix acts naturally on its eigenvectors. Since matrix polynomials respect the structure of eigenspaces, applying  $g(A)$  to an eigenvector  $\mathbf{v}$  results in  $g(A)\mathbf{v} = g(\lambda)\mathbf{v}$ . Since we already assumed  $g(A)\mathbf{v} = c\mathbf{v}$ , comparing the two equations gives  $c\mathbf{v} = g(\lambda)\mathbf{v}$ . Since  $\mathbf{v} \neq \mathbf{0}$ , we must have  $c = g(\lambda)$ .

Suppose  $\lambda$  is an eigenvalue of  $A$  and  $c = g(\lambda)$ . Since  $\lambda$  is an eigenvalue of  $A$ , there exists a nonzero vector  $\mathbf{v} \in \mathbb{C}^n$  such that  $A\mathbf{v} = \lambda\mathbf{v}$ . Applying  $g(A)$  to  $\mathbf{v}$ , we compute  $g(A)\mathbf{v} = g(\lambda)\mathbf{v} = c\mathbf{v}$ . This shows that  $\mathbf{v}$  is an eigenvector of  $g(A)$  with eigenvalue  $c$ , meaning that  $c$  is an eigenvalue of  $g(A)$ .

Therefore,  $c$  is an eigenvalue of  $g(A)$  if and only if  $c = g(\lambda)$  for some eigenvalue  $\lambda$  of  $A$ .  $\square$

**Problem 6.** Suppose  $V = W_1 \oplus W_2$ . Prove that for any  $\mathbf{v} \in V$ , there exists a unique pair of vectors  $\mathbf{w}_1 \in W_1$  and  $\mathbf{w}_2 \in W_2$  such that  $\mathbf{v} = \mathbf{w}_1 + \mathbf{w}_2$ .

*Solution.* To prove the statement, we must show that for every  $\mathbf{v} \in V$ , there exists a unique pair  $(\mathbf{w}_1, \mathbf{w}_2)$ , with  $\mathbf{w}_1 \in W_1$  and  $\mathbf{w}_2 \in W_2$  such that  $\mathbf{v} = \mathbf{w}_1 + \mathbf{w}_2$ .

Since  $V = W_1 \oplus W_2$ , by definition of the direct sum, every vector  $\mathbf{v} \in V$  can be written as a sum of two vectors  $\mathbf{v} = \mathbf{w}_1 + \mathbf{w}_2$ , where  $\mathbf{w}_1 \in W_1$  and  $\mathbf{w}_2 \in W_2$ . Thus, such a decomposition always exists.

Suppose there exist two different representations  $\mathbf{v} = \mathbf{w}_1 + \mathbf{w}_2$  and  $\mathbf{v} = \mathbf{w}'_1 + \mathbf{w}'_2$ , where  $\mathbf{w}_1, \mathbf{w}'_1 \in W_1$  and  $\mathbf{w}_2, \mathbf{w}'_2 \in W_2$ . Then we have  $\mathbf{w}_1 + \mathbf{w}_2 = \mathbf{w}'_1 + \mathbf{w}'_2$ . Rearranging gives us  $\mathbf{w}_1 - \mathbf{w}'_1 = \mathbf{w}_2 - \mathbf{w}'_2$ . Since  $\mathbf{w}_1 - \mathbf{w}'_1 \in W_1$  and  $\mathbf{w}_2 - \mathbf{w}'_2 \in W_2$  and because  $W_1 \cap W_2 = \{\mathbf{0}\}$ , we must have  $\mathbf{w}_1 - \mathbf{w}'_1 = \mathbf{0}$  and  $\mathbf{w}_2 - \mathbf{w}'_2 = \mathbf{0}$ . That is,  $\mathbf{w}_1 = \mathbf{w}'_1$  and  $\mathbf{w}_2 = \mathbf{w}'_2$ . Thus, the pair  $(\mathbf{w}_1, \mathbf{w}_2)$  is unique.  $\square$

**Problem 7.** True or False. (No explanation needed.)

- (i) Let  $A \in \mathbb{C}^{n \times n}$ , then  $A$  has exactly  $n$  eigenvalues (counting the multiplicities).
- (ii) Let  $T : V \rightarrow V$  be a linear transformation, where  $\dim(V) = n$ . Then  $T$  is diagonalizable if and only if  $T$  has  $n$  distinct eigenvalues.
- (iii) Similar matrices always have the same eigenvalues.
- (iv) Similar matrices always have the same eigenvectors.
- (v) The sum of two eigenvectors of a linear transformation  $T$  is always an eigenvector of  $T$ .
- (vi) If  $\lambda_1$  and  $\lambda_2$  are distinct eigenvalues of a linear transformation, then  $E_{\lambda_1} \cap E_{\lambda_2} = \{\mathbf{0}\}$ .
- (vii) A linear transformation  $T$  on a finite-dimensional vector space is diagonalizable if and only if the algebraic multiplicity of each eigenvalue  $\lambda$  equals to its geometric multiplicity.
- (viii) Suppose  $W_1, W_2, \dots, W_m \subset V$  are subspaces. Then  $W_1 + W_2 + \dots + W_m$  is a direct sum if  $W_i \cap W_j = \{\mathbf{0}\}$  for any  $i \neq j$ .

*Solution to (i).* True.  $\square$

*Solution to (ii).* False.  $\square$

*Solution to (iii).* True.  $\square$

*Solution to (iv).* False.  $\square$

*Solution to (v).* False.  $\square$

*Solution to (vi).* True.  $\square$

*Solution to (vii).* True.  $\square$

*Solution to (viii).* False.  $\square$