

# 1 Exercises to Prepare for Test 3

**Q01.** Let  $C \in M_{n \times n}(\mathbb{F})$  be invertible, and let  $\mathbf{v}_1, \dots, \mathbf{v}_k \in \mathbb{F}^n$ . Prove that if  $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$  is linearly independent, then so is  $\{C\mathbf{v}_1, \dots, C\mathbf{v}_k\}$ .

*Proof* (from Rajvi). Suppose that  $\sum a_i C\mathbf{v}_i = \mathbf{0}$ . We must show that all  $a_i = 0$ .

By the linearity of matrix multiplication,  $\sum a_i C\mathbf{v}_i = C \sum a_i \mathbf{v}_i$ . However, since  $C$  is invertible, we have  $\sum a_i \mathbf{v}_i = \mathbf{0}$ . Since  $\{\mathbf{v}_i\}$  is linearly independent, this only occurs if all  $a_i = 0$ .  $\square$

*Proof* (more complicated). Proceed by the contrapositive.

Suppose that  $\{C\mathbf{v}_i\}$  is linearly dependent. Then,  $\sum a_i C\mathbf{v}_i = \mathbf{0}$  for some non-zero  $a_i$ . By linearity,  $C \sum a_i \mathbf{v}_i = \mathbf{0}$ . Since  $C$  is invertible,  $\sum a_i \mathbf{v}_i = \mathbf{0}$ . This is exactly what it means for  $\{\mathbf{v}_i\}$  to be linearly dependent.  $\square$

**Q02.** Let  $L : \mathbb{F}^n \rightarrow \mathbb{F}^m$  be a linear mapping, and let  $\mathbf{v}_1, \dots, \mathbf{v}_k \in \mathbb{F}^n$ .

- (a) Prove or disprove: if  $L$  is one-to-one and  $\{L(\mathbf{v}_1), \dots, L(\mathbf{v}_k)\}$  is linearly independent, then so is  $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ .

*Proof.* Proceed by the contrapositive. Suppose that  $\{\mathbf{v}_i\}$  is linearly dependent, so  $\sum c_i \mathbf{v}_i = \mathbf{0}$  for non-zero  $c_i$ . Now, if we apply  $L$  to both sides,  $\sum c_i L(\mathbf{v}_i) = L(\mathbf{0})$  by linearity. But  $L(\mathbf{0}) = \mathbf{0}$ , so we are done.  $\square$

- (b) Prove or disprove: if  $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$  is linearly independent, then so is  $\{L(\mathbf{v}_1), \dots, L(\mathbf{v}_k)\}$ .

*Solution.* For a counterexample, define  $L$  by the mapping  $\mathbf{x} \mapsto \mathbf{0}$ .

Then,  $L(\mathbf{v}_1) = \mathbf{0}$  so any set containing it is linearly dependent.  $\square$

**Q03.** Let  $A \in M_{n \times n}(\mathbb{F})$ . We say that  $A$  is nilpotent if there exists a positive integer  $n$  such that  $A^n = \mathcal{O}_{n \times n}$ . Prove that  $\lambda = 0$  is the only eigenvalue of  $A$ .

*Proof.* Start by taking the determinant on both sides. Then,  $\det(A^n) = \det(A)^n = \det(\mathbf{0}) = 0$ . Therefore,  $\det(A) = 0$ .

Then, we have a non-trivial solution  $\mathbf{x}$  to  $A\mathbf{x} = \mathbf{0}$ . But this is just  $A\mathbf{x} = \lambda\mathbf{x}$  for  $\lambda = 0$ . Therefore, 0 is an eigenvalue of  $A$ .

Now, we prove uniqueness. Suppose that  $A\mathbf{x} = \lambda\mathbf{x}$  for arbitrary  $\lambda$  and non-zero  $\mathbf{x}$ . Multiply on the left by  $A^{n-1}$ . Then,  $A^n\mathbf{x} = A^{n-1}\lambda\mathbf{x}$ . But this expands as  $A^n\mathbf{x} = \lambda^n\mathbf{x}$ . Since  $A^n = \mathcal{O}$ , we have  $\mathbf{0} = \lambda^n\mathbf{x}$ . But  $\mathbf{x}$  is non-zero, so  $\lambda^n = 0$  and  $\lambda = 0$ .

Therefore, the only eigenvalue of  $A$  is 0.  $\square$

**Q04.** Let  $\mathbf{v}_1, \dots, \mathbf{v}_n \in \mathbb{F}^n$ , and let  $c_1, \dots, c_n \in \mathbb{F}$  be non-zero scalars. Prove that if  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $\mathbb{F}^n$ , then so is  $\{c_1\mathbf{v}_1, \dots, c_n\mathbf{v}_n\}$ .

*Proof.* We must show that  $\{c_i\mathbf{v}_i\}$  is both spanning and linearly independent.

For spanning, notice that it follows trivially from the definition that multiplying a term of a linear combination by a non-zero scalar does not change the span.

Let  $B = \{\mathbf{v}_i\}$  and let  $[C]_B = \text{diag}(c_i)$ . Then,  $C\mathbf{v}_i = c_i\mathbf{v}_i$  and  $C$  is invertible since it is diagonal. But by Q01,  $\{C\mathbf{v}_i\}$  is linearly independent. Therefore, since  $\{c_i\mathbf{v}_i\}$  is spanning and linearly independent, it is a basis.  $\square$

**Q05.** Let  $\mathbf{v}_1, \dots, \mathbf{v}_n \in \mathbb{F}^n$ , and let  $B$  be a basis of  $\mathbb{F}^n$ . Prove that  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $\mathbb{F}^n$  if and only if  $\{[\mathbf{v}_1]_B, \dots, [\mathbf{v}_n]_B\}$  is a basis of  $\mathbb{F}^n$ .

*Proof.* Notice that the proof goes in both directions if we consider bases generally. Again, we must show spanning and linear independence.

Since  $\{\mathbf{v}_i\}$  is a basis, the matrix  $(\mathbf{v}_i)$  is invertible. Then,  $([\mathbf{v}_i]_B) = [(\mathbf{v}_i)]_B = {}_B[I]_S(\mathbf{v}_i)_S[I]_B$  must also be invertible as the product of invertible matrices. Therefore,  $\{[\mathbf{v}_i]_B\}$  is invertible and therefore spanning.

Proceed as in Q04 to show linear independence with  $C = {}_B[I]_S$ .

Conversely, consider when  $B = S$  and  $S = B$ . □

**Q06.** Let  $\mathbf{v}_1, \dots, \mathbf{v}_n \in \mathbb{F}^n$ . Prove that if for every vector  $\mathbf{x} \in \mathbb{F}^n$ , there exist unique scalars  $c_1, \dots, c_n \in \mathbb{F}$  such that  $\mathbf{x} = c_1\mathbf{v}_1 + \dots + c_n\mathbf{v}_n$ , then  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis of  $\mathbb{F}^n$ .

*Proof.* We must prove spanning and linear independence. Spanning follows immediately from the hypothesis by definition.

By Lemma 17C.11,  $\{\mathbf{v}_i\}$  is linearly independent since there are  $n$  vectors.

Therefore, it is a basis. □

**Q07.** Find all real numbers  $a$  and  $b$  such that  $\text{Span} \left( \left\{ \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ a \\ 2 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ b \end{pmatrix} \right\} \right) \neq \mathbb{R}^3$ .

*Solution.* Consider  $A = \begin{pmatrix} 1 & 1 & 0 \\ 2 & a & 1 \\ 1 & 2 & b \end{pmatrix}$ . We consider when  $\text{Col}(A) \neq \mathbb{R}^3$ .

By the Rank-Nullity Theorem, we must find when  $N(A) \neq \{\mathbf{0}\}$ . This occurs only when  $\det(A) = 0$ . Expanding the determinant,  $ab - 2b - 1 = 0$ , so  $b = \frac{1}{a-2}$ .

Therefore, for all  $(a, b) \in \{(k, \frac{1}{k-2}) : k \in \mathbb{R} \setminus \{2\}\}$ ,  $\text{Col}(A) \neq \mathbb{R}^3$ . □

**Q08.** Let  $V = \left\{ \begin{pmatrix} a^2 \\ b \\ b \end{pmatrix} : a, b \in \mathbb{R} \right\}$  be a subset of  $\mathbb{F}^3$ . Prove or disprove:

(a) If  $\mathbb{F} = \mathbb{R}$ , then  $V$  is a subspace of  $\mathbb{F}^3$ .

(b) If  $\mathbb{F} = \mathbb{C}$ , then  $V$  is a subspace of  $\mathbb{F}^3$ .

*Solution.* Notice that  $V$  is defined as a subset of  $\mathbb{R}^3$  since the parameters are in  $\mathbb{R}$ . Then, we know  $\mathbf{x} = (1, 0, 0)^T \in V$  with  $a = 1$  and  $b = 0$ .

However,  $-2\mathbf{x} = (-2, 0, 0)^T \notin V$  because there exists no  $a \in \mathbb{R}$  such that  $a^2 = -2$ . Therefore,  $V$  is not closed under scalar multiplication.

Since  $-2 \in \mathbb{R}$  and  $-2 \in \mathbb{C}$ ,  $V$  is neither a subspace of  $\mathbb{R}^3$  nor  $\mathbb{C}^3$ . □

**Q09.** We call a square matrix  $A$  idempotent if  $A^2 = A$ . Prove that if  $A$  is idempotent, then so is  $I - A$ . Is the converse of this statement true? Explain why or why not.

*Proof.* Suppose that  $A^2 = A$ . Then,  $(I - A)^2 = (I - A)(I - A) = I^2 - AI - IA + A^2 = I - 2A + A^2 = I - A$  by properties of the identity matrix and the distributivity of matrix multiplication. Therefore,  $I - A$  is idempotent.

Suppose conversely that  $(I - A)^2 = I - A$ . Then,  $I - 2A + A^2 = I - A$  as above, but then  $-A + A^2 = \mathbf{0}$ . It follows  $A = A^2$  and  $A$  is idempotent. □

**Q10.** Let  $A, B \in M_{n \times n}(\mathbb{F})$ .

- (a) Prove or disprove: if  $\mathbf{v}$  is an eigenvector of both  $A$  and  $B$ , then it is an eigenvector of both  $AB$  and  $BA$ .

*Proof.* Suppose  $A\mathbf{v} = \lambda_A\mathbf{v}$  and  $B\mathbf{v} = \lambda_B\mathbf{v}$ .

If we multiply the first equation by  $B$ , we have  $BA\mathbf{v} = B\lambda_A\mathbf{v} = \lambda_AB\mathbf{v} = \lambda_A\lambda_B\mathbf{v}$ .

If we instead multiply the second by  $A$ , we have  $AB\mathbf{v} = A\lambda_B\mathbf{v} = \lambda_BA\mathbf{v} = \lambda_B\lambda_A\mathbf{v}$ .

Therefore,  $\mathbf{v}$  is an eigenvector of  $AB$  and  $BA$ .  $\square$

- (b) Prove or disprove: if  $\lambda$  is an eigenvalue of both  $A$  and  $B$ , then it is an eigenvalue of both  $AB$  and  $BA$ .

*Solution.* We consider for a counterexample  $A = B = \begin{pmatrix} 2 & 0 \\ 0 & 1 \end{pmatrix}$ . Then,  $\lambda = 2$  is an eigenvalue of  $A$  and  $B$ .

However,  $AB = BA = \begin{pmatrix} 4 & 0 \\ 0 & 1 \end{pmatrix}$  and  $\lambda = 2$  is not an eigenvalue.  $\square$

## 2 Exercises to Prepare for the Exam

Sourced from Piazza @4051.

**Q01.** We say that a subset  $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$  of  $\mathbb{C}^n$  is orthogonal if  $\langle \mathbf{v}_i, \mathbf{v}_j \rangle = 0$  for all  $i \neq j$ . Prove that  $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$  is orthogonal if and only if  $\{\overline{\mathbf{v}}_1, \dots, \overline{\mathbf{v}}_k\}$  is orthogonal.

*Proof.* Suppose that  $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$  is orthogonal. Then,  $\langle \mathbf{v}_i, \mathbf{v}_j \rangle = 0$  for all  $i \neq j$ .

But  $\langle \overline{\mathbf{v}}_i, \overline{\mathbf{v}}_j \rangle = \sum \overline{v_{ii}} v_{ji} = \overline{\sum v_{ii} \overline{v_{ji}}} = \overline{\langle \mathbf{v}_i, \mathbf{v}_j \rangle} = \overline{0} = 0$ . Thus,  $\{\overline{\mathbf{v}}_1, \dots, \overline{\mathbf{v}}_k\}$  is orthogonal.

Conversely, notice that  $\overline{\overline{\mathbf{v}}_i} = \mathbf{v}_i$  for any vector.  $\square$

**Q02.** Let  $A \in M_{n \times n}$  be diagonalizable with not necessarily distinct eigenvalues  $\lambda_1, \dots, \lambda_n$ . Prove that  $B = A - \lambda_1 I$  is diagonalizable. What are the eigenvalues of  $B$ ?

*Proof.* Let  $\mathcal{B} = \{\mathbf{v}_i\}$  be the eigenvectors of  $A$  with eigenvalues  $\lambda_i$ .

Then,  $B\mathbf{v}_i = (A - \lambda_1 I)\mathbf{v}_i = A\mathbf{v}_i - \lambda_1 \mathbf{v}_i = (\lambda_i - \lambda_1)\mathbf{v}_i$  for all  $i$ .

Therefore,  $[B]_{\mathcal{B}} = \text{diag}(0, \lambda_2 - \lambda_1, \dots, \lambda_n - \lambda_1)$  so  $B$  is diagonalizable. Notice that because  $|\mathcal{B}| = n$ , these are the only eigenvalues of  $B$ .  $\square$

**Q03.** Let  $B \in M_{n \times n}$ . Prove or disprove: if  $B^2 \mathbf{x} = \mathbf{0}$  for some  $\mathbf{x} \neq \mathbf{0}$ , then  $B$  is not invertible.

*Proof.* Recall that by the Invertible Matrix Theorem,  $B^2 \mathbf{x} = \mathbf{0}$  for some  $\mathbf{x} \neq \mathbf{0}$  iff  $B^2$  is not invertible. Proceed by the contrapositive. Suppose that  $B$  is invertible. Then,  $B^2 = BB$  must be invertible as the product of invertible matrices.  $\square$

**Q04.** Let  $A \in M_{n \times n}$  be diagonal with not necessarily distinct non-zero eigenvalues  $\lambda_1, \dots, \lambda_k$ . Prove that there exist eigenvectors  $\mathbf{v}_1, \dots, \mathbf{v}_k$  of  $A$  such that the set  $B = \{\mathbf{v}_1, \dots, \mathbf{v}_k\}$  is a basis of  $\text{Col}(A)$ .

*Proof.* Notice that the diagonal entries of  $A$  are either the  $k$  non-zero eigenvalues or zero. Therefore,  $\text{rank}(A) = \dim(\text{Col}(A)) = k$ , since all other columns are zero.

Without loss of generality, suppose that  $\lambda_i$  is in the  $i$ th column of  $A$ . Then,  $A\mathbf{e}_i = \lambda_i \mathbf{e}_i$  for each standard basis vector with  $i \leq k$ . Notice also that  $\mathbf{e}_i$  is an eigenvector of  $A$ . Also, since  $\lambda_i$  is non-zero,  $A(\lambda_i^{-1} \mathbf{e}_i) = \mathbf{e}_i$  so we have  $\mathbf{e}_i \in \text{Col}(A)$ .

Now, let  $\mathbf{v}_i = \mathbf{e}_i$  for each  $i \leq k$ . The set  $B$  is linearly independent as a subset of  $S$  with  $|B| = k$  and  $B \subset \text{Col}(A)$ . By Lemma 17C.11, it is a basis for  $\text{Col}(A)$ .  $\square$

**Q05.** Let  $T : \mathbb{F}^n \rightarrow \mathbb{F}^n$  be a linear transformation. Prove or disprove: if  $T$  is not invertible, then there exists a basis  $B$  for  $\mathbb{F}^n$  for which the matrix  $[T]_B$  has a column of zeroes.

*Proof.* Suppose  $T$  is not invertible. Recall that by the Invertible Matrix Theorem, the nullity of  $T$  is at least 1. Therefore, there exists a basis  $B_1$  of  $N(T)$  with  $|B_1| \geq 1$ .

Then, by the Replacement Theorem, there exists a basis  $B$  for  $\mathbb{F}^n$  consisting of  $B_1$  and some of the standard basis vectors. Now, let  $\mathbf{v} \in B_1$ . Then,  $T\mathbf{v} = \mathbf{0}$ , so the column of  $[T]_B$  corresponding to  $\mathbf{v}$  has all zero entries.  $\square$

**Q06.** Let  $T$  be an invertible linear operator on  $\mathbb{F}^n$ .

(a) Show that if  $\lambda$  is an eigenvalue of  $T$  then  $\lambda^{-1}$  is an eigenvalue of  $T^{-1}$ .

*Proof.* Let  $T(\mathbf{v}) = \lambda \mathbf{v}$ . Then,  $\mathbf{v} = T^{-1}(T(\mathbf{v})) = \lambda T^{-1}(\mathbf{v})$  and so  $T^{-1}(\mathbf{v}) = \lambda^{-1} \mathbf{v}$ .  $\square$

- (b) Show that the eigenspace of  $T$  corresponding to  $\lambda$  is equal to the eigenspace of  $T^{-1}$  corresponding to  $\lambda^{-1}$ .

*Proof.* Notice that in (a) the same vector  $\mathbf{v}$  was the eigenvector of  $T$  corresponding to  $\lambda$  and of  $T^{-1}$  corresponding to  $\lambda^{-1}$ .

Therefore,  $E_\lambda(T) = \text{Span}(\{\mathbf{v}, \dots\}) = E_{\lambda^{-1}}(T^{-1})$ .  $\square$

- (c) Prove or disprove:  $T$  is diagonalizable iff  $T^{-1}$  is diagonalizable.

*Proof.* Recall that  $T$  is diagonalizable iff  $T$  has  $n$  linearly independent eigenvectors. However, the eigenvectors of  $T$  and  $T^{-1}$  are the same. Therefore,  $T^{-1}$  is diagonalizable iff  $T$  has  $n$  linearly independent eigenvectors iff  $T$  is diagonalizable.  $\square$

**Q07.** Suppose that  $A \in M_{n \times n}$  has two distinct eigenvalues  $\lambda_1$  and  $\lambda_2$ . Show that if the geometric multiplicity of one of the eigenvalues is  $n - 1$ , then  $A$  must be diagonalizable. Is the converse true?

*Proof.* Suppose  $g_{\lambda_1} = n - 1$ . Then, by definition, a basis  $B_1$  for  $E_{\lambda_1}$  has  $n - 1$  vectors.

Now, let  $B_2$  be a basis for  $E_{\lambda_2}$ . By Lemma 19B.6,  $E_{\lambda_1} \cup E_{\lambda_2}$  is linearly independent. But if  $\dim(B_2) > 1$ , then  $\dim(E_{\lambda_1} \cup E_{\lambda_2}) > n$ . Therefore,  $\dim(B_2) = g_{\lambda_2} = 1$ .

We now have  $n$  linearly independent eigenvectors, so  $A$  is diagonalizable.

The converse is not true. Consider  $A = \text{diag}(1, 2, 3)$ . No eigenvalue has multiplicity 2.  $\square$

**Q08.** Suppose  $A \in M_{n \times n}$  is similar to the upper triangular matrix 
$$\begin{pmatrix} 1 & 2 & 3 & \cdots & n \\ 0 & 2 & 3 & \cdots & n \\ 0 & 0 & 3 & \cdots & n \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 & n \end{pmatrix}.$$

Show that  $A$  is diagonalizable.

*Solution.* Let  $B$  be the given upper triangular matrix. Then, we can evaluate along the diagonal  $\Delta_B(t) = (1 - t)(2 - t)(3 - t) \cdots (n - t)$ .

Since  $A$  is similar to  $B$ , it has the same characteristic polynomial. That is,  $A$  also has eigenvalues  $\lambda = 1, \dots, n$ . Since there are  $n$  distinct eigenvalues,  $A$  is diagonalizable.  $\square$

**Q09.** Let  $X = (\mathbf{v}_1, \dots, \mathbf{v}_k)$  and  $Y = (\mathbf{u}_1, \dots, \mathbf{u}_k)$  be in  $M_{n \times k}$ .

Prove that if  $\text{Col}(X) \subseteq \text{Col}(Y)$ , then there exists  $A \in M_{k \times k}$  such that  $X = YA$ .

*Proof.*  $\square$

**Q10.** Suppose that  $L : \mathbb{F}^4 \rightarrow \mathbb{F}^7$  is linear. Prove that  $R(L) \neq \mathbb{F}^7$ .

*Proof.* Suppose  $R(L) = \mathbb{F}^7$ . Then,  $\dim(\text{Col}(L)) = 7$ . However, there are only four columns of  $[L]_S$ . Therefore,  $\dim(\text{Col}(L)) \leq 4$ , so by contradiction,  $R(L) \neq \mathbb{F}^7$ .  $\square$

**Q11.** Let  $A \in M_{3 \times 3}(\mathbb{R})$  with three distinct, real, non-negative eigenvalues.

Prove that there exists  $B \in M_{3 \times 3}(\mathbb{R})$  such that  $B^2 = A$ .

*Proof.* As the eigenvalues are distinct, we may diagonalize  $A$ . Let  $\mathcal{B}$  be the eigenbasis for  $A$  such that  $[A]_{\mathcal{B}} = \text{diag}(\lambda_1, \lambda_2, \lambda_3)$ . Now, since all eigenvalues are non-negative, let  $[B]_{\mathcal{B}} = \text{diag}(\sqrt{\lambda_1}, \sqrt{\lambda_2}, \sqrt{\lambda_3})$ .

Therefore,  $B = {}_S[I]_{\mathcal{B}} \text{diag}(\sqrt{\lambda_1}, \sqrt{\lambda_2}, \sqrt{\lambda_3})_{\mathcal{B}}[I]_S \in M_{3 \times 3}(\mathbb{R})$  such that  $B^2 = A$ .  $\square$

**Q12.** Let  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  be a basis of  $\mathbb{F}^n$ .

Prove that for any basis  $B$  for  $\mathbb{F}^n$ ,  $\{[\mathbf{v}_1]_B, \dots, [\mathbf{v}_n]_B\}$  is a basis for  $\mathbb{F}^n$ .

*Proof.* This is a specific case of Question 1.05 above. □

**Q13.** Let  $A, B \in M_{n \times n}$ , with  $AB = BA$ . Suppose that every eigenvalue of  $A$  has algebraic multiplicity 1. Prove that every eigenvector of  $A$  is an eigenvector of  $B$ .

*Proof.* Let  $A\mathbf{v}_i = \lambda_i\mathbf{v}_i$ . Now, because  $AB = BA$ , we can write  $AB\mathbf{v}_i = BA\mathbf{v}_i = \lambda_i(B\mathbf{v}_i)$ . But this means  $B\mathbf{v}_i$  is an eigenvector for  $A$  with eigenvalue  $\lambda_i$ , that is,  $B\mathbf{v}_i \in E_{\lambda_i}$ .

By Lemma 19B.5, if  $a_{\lambda_i} = 1$ , then  $g_{\lambda_i} = 1$ . Then,  $\dim(E_{\lambda_i}) = 1$ , so  $E_{\lambda_i} = \text{Span}(\{\mathbf{v}_i\})$ . Therefore,  $B\mathbf{v}_i = k\mathbf{v}_i$  by definition of the span. But this is exactly what it means for  $\mathbf{v}_i$  to be an eigenvector of  $B$ . □

**Q14.** Let  $A, B \in M_{4 \times 4}$  with  $A = (\mathbf{a}_1, \mathbf{a}_2, \mathbf{a}_3, \mathbf{a}_4)$  and  $B = (\mathbf{a}_4, \mathbf{a}_3, \mathbf{a}_2, \mathbf{a}_1)$ .

Prove that  $A - B$  is not invertible.

*Proof.* Notice that we have  $A - B = (\mathbf{a}_1 - \mathbf{a}_4, \mathbf{a}_2 - \mathbf{a}_3, \mathbf{a}_3 - \mathbf{a}_2, \mathbf{a}_4 - \mathbf{a}_1)$ . But if we perform  $C_4 + C_1$  and  $C_3 + C_2$  we have  $A - B = (\mathbf{a}_1 - \mathbf{a}_4, \mathbf{a}_2 - \mathbf{a}_3, 0, 0)$ . Therefore,  $\text{rank}(A - B) \leq 2$  since  $\dim(\text{Col}(A - B)) = \dim(\text{Row}(A - B))$ . It follows  $A - B$  is not invertible. □

**Q15.** Let  $\mathbf{n} = (i, 1, 0, 1 + i)^T$ . Compute a basis of a subspace  $S = \{\mathbf{x} \in \mathbb{C}^4 : \langle \mathbf{x}, \mathbf{n} \rangle = 0\}$ . What is  $\dim(S)$ ?

*Solution.* Let  $\mathbf{x} = (a, b, c, d)^T \in S$ . Then,  $\langle \mathbf{x}, \mathbf{n} \rangle = a(-i) + b + d(1 - i) = 0$ . Equivalently,  $b = ia - (1 - i)d$  for arbitrary  $a$  and  $d$ .

Let  $a = s$ ,  $c = t$ , and  $d = u$ . Then,  $S = \{(s, is + (-1 + i)u, t, u)^T : s, t, u \in \mathbb{C}\}$ .

But this is equivalently  $S = \text{Span} \left( \left\{ \begin{pmatrix} 1 \\ i \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ -1 + i \\ 0 \\ 1 \end{pmatrix} \right\} \right)$ .

The set is linearly independent, so it forms a basis for  $S$ . Therefore,  $\dim(S) = 3$ . □

**Q16.** Let  $A \in M_{n \times n}$  be non-invertible. Prove that the eigenspace of  $A$  corresponding to the eigenvalue  $\lambda = 0$  is equal to the nullspace of  $A$ .

*Proof.* By the Invertible Matrix Theorem, since  $A$  is singular,  $E_0$  exists. Now, notice that  $E_0$  is defined by  $N(A - 0I) = N(A)$ . □

**Q17.** Find the projection of  $\mathbf{v} = \begin{pmatrix} -2 \\ 0 \\ 3 \end{pmatrix}$  onto the plane  $S = \text{Span} \left( \left\{ \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix} \right\} \right)$ .

*Solution.* Who needs fancy matrices? We must take  $\mathbf{v} - \text{Proj}_{\mathbf{n}}(\mathbf{v})$  for a normal vector  $\mathbf{n}$ . Note that we have a trivial  $\mathbf{n} = (2, 1, 1)^T \times (1, 1, -1)^T = (-2, 3, 1)^T$ . Then, we can calculate that  $\text{Proj}_{\mathbf{n}}(\mathbf{v}) = \frac{1}{2}(-2, 3, 1)^T$ .

Therefore, the projection is  $\frac{1}{2}(-2, -3, 5)^T$ . □