

**This homework is due at 11 PM on February 6th, 2026.**

**Submission Format:** Your homework submission should consist of a single PDF file that contains all of your answers (any handwritten answers should be scanned).

**1. Proof of the Fundamental Theorem of Linear Algebra**

In this question, we will prove the fundamental theorem of linear algebra. For any  $A \in \mathbb{R}^{m \times n}$ , let  $\mathcal{N}(A)$ ,  $\mathcal{R}(A)$ , and  $\text{rank}(A)$  denote the null space, range and rank of  $A$  respectively. For any subspace,  $\mathcal{S}$  with dimension,  $\dim(\mathcal{S})$ , let  $\mathcal{S}^\perp$  denote its the subspace orthogonal to  $\mathcal{S}$ . The fundamental theorem of linear algebra states that,

$$\mathcal{N}(A) \oplus \mathcal{R}(A^\top) = \mathbb{R}^n. \quad (1)$$

The proof technique we employ will first show that,

$$\mathcal{N}(A) = \mathcal{R}(A^\top)^\perp. \quad (2)$$

Then we will prove that we can find orthonormal vectors  $\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n$  such that  $\mathcal{N}(A) = \text{span}(\vec{e}_1, \vec{e}_2, \dots, \vec{e}_\ell)$  and  $\mathcal{R}(A^\top) = \text{span}(\vec{e}_{\ell+1}, \vec{e}_{\ell+2}, \dots, \vec{e}_n)$ . As a corollary we get the rank-nullity theorem:

$$\dim(\mathcal{N}(A)) + \text{rank}(A) = n. \quad (3)$$

- (a) First, show that  $\mathcal{N}(A) \subseteq \mathcal{R}(A^\top)^\perp$ .

*HINT: Consider  $\vec{u}$  in  $\mathcal{N}(A)$ ,  $\vec{v} \in \mathcal{R}(A^\top)$  and show that  $\vec{u}^\top \vec{v} = 0$ .*

- (b) Now show that:  $\mathcal{R}(A^\top)^\perp \subseteq \mathcal{N}(A)$ .

*HINT: Show that any vector  $\vec{v}$  that is orthogonal to all vectors in the range of  $A^\top$  satisfies  $A\vec{v} = 0$ . To do this, consider  $\vec{v} \in \mathcal{R}(A^\top)^\perp$  and what it implies for  $\vec{v}^\top A^\top$ .*

- (c) Note that we could apply the orthogonal decomposition theorem (theorem 19 in the course reader) at this point to complete the proof. However, instead we'll work through how to re-derive that result directly. Let  $\dim(\mathcal{N}(A)) = \ell$  and let  $\vec{e}_1, \dots, \vec{e}_\ell$  be an orthonormal basis for  $\mathcal{N}(A)$ . Consider an extension of the basis to an orthonormal basis,  $\vec{e}_1, \dots, \vec{e}_n$  for  $\mathbb{R}^n$ . We will prove that  $\vec{e}_{\ell+1}, \dots, \vec{e}_n$  form a basis for  $\mathcal{R}(A^\top)$  and as a consequence, the dimension of  $\mathcal{R}(A^\top)$  is  $n - \ell$ .

- i. Show that  $\mathcal{R}(A^\top)$  lies in the span of  $\vec{e}_{\ell+1}, \dots, \vec{e}_n$ .

*HINT: Express any vector  $\vec{u} \in \mathcal{R}(A^\top)$  as  $\vec{u} = \sum_{i=1}^n \alpha_i \vec{e}_i$ . What are the values of  $\alpha_i$ ?*

- ii. From part (i) we know that  $\mathcal{R}(A^\top) \subseteq \text{span}(\vec{e}_{\ell+1}, \dots, \vec{e}_n)$ , but we want something stronger. Show that in fact  $\mathcal{R}(A^\top) = \text{span}(\vec{e}_{\ell+1}, \dots, \vec{e}_n)$ .

*HINT: First, prove  $\dim(\mathcal{R}(A^\top)) = \dim(\text{span}(\vec{e}_{\ell+1}, \dots, \vec{e}_n)) = n - \ell$  by contradiction. Assume  $\dim(\mathcal{R}(A^\top)) = k < n - \ell$ , show that a vector  $\vec{u} \in \text{span}(\vec{e}_{\ell+1}, \dots, \vec{e}_n)$  and  $\vec{u} \notin \mathcal{R}(A^\top)$  cannot exist.*

*Specifically, let  $\vec{f}_1, \vec{f}_2, \dots, \vec{f}_k$  be an orthonormal basis for  $\mathcal{R}(A^\top)$ , we can find non-zero  $\vec{u}_\perp = \vec{u} - \sum_{i=1}^k (\vec{f}_i^\top \vec{u}) \vec{f}_i$  that is orthogonal to  $\mathcal{R}(A^\top)$ . Does  $\vec{u}_\perp$  lie in  $\mathcal{N}(A)$ ? Does  $\vec{u}_\perp$  also lie in  $\text{span}(\vec{e}_{\ell+1}, \dots, \vec{e}_n)$ ? Does this lead to a contradiction? Think of  $n - \ell = 3$  and  $k = 2$  for visualization.*

*HINT: Second, you can use without proof the fact that for two subspaces,  $\mathcal{S}_1$  and  $\mathcal{S}_2$ , if  $\mathcal{S}_1 \subseteq \mathcal{S}_2$  and  $\dim(\mathcal{S}_1) = \dim(\mathcal{S}_2)$  then  $\mathcal{S}_1 = \mathcal{S}_2$ .*

(d) Using part (c) argue why  $\mathcal{N}(A) \oplus \mathcal{R}(A^\top) = \mathbb{R}^n$  and why the rank nullity theorem holds.

## 2. Eigenvalues of Symmetric Matrices

Let  $A \in \mathbb{S}^n$  (i.e., the set of  $n \times n$  real symmetric matrices) with eigenvalues  $\lambda_i$ . Prove that all of the eigenvalues of  $A$  are real (i.e. that  $\lambda_i \in \mathbb{R}$  for each  $i$ ).

*HINT: Consider the quantity  $(Av)^*v$  for eigenvector  $v$  where  $*$  denotes the conjugate transpose. Note that this is the Hermitian inner product between  $Av$  and  $v$ .*

*NOTE:* This exercise is part of the proof of the spectral theorem.

**3. Distinct Eigenvalues, Orthogonal Eigenspaces**

Let  $A \in \mathbb{S}^n$  (i.e. the set of  $n \times n$  real symmetric matrices) and  $(\lambda_1, \vec{u}_1), (\lambda_2, \vec{u}_2), \lambda_1 \neq \lambda_2$  be distinct eigen-pairs of  $A$ . Show that  $\vec{u}_1^\top \vec{u}_2 = 0$ , i.e., eigenspaces corresponding to distinct eigenvalues are mutually orthogonal.

*HINT: First try to prove that  $\lambda_1 \vec{u}_1^\top \vec{u}_2 = \lambda_2 \vec{u}_1^\top \vec{u}_2$ , then show that this implies  $\vec{u}_1^\top \vec{u}_2 = 0$ .*

*NOTE:* This exercise is part of the proof of the spectral theorem.

#### 4. Gram Schmidt

Any set of  $n$  linearly independent vectors in  $\mathbb{R}^n$  could be used as a basis for  $\mathbb{R}^n$ . However, certain bases could be more suitable for certain operations than others. For example, an orthonormal basis could facilitate solving linear equations.

- (a) Given a matrix  $A \in \mathbb{R}^{n \times n}$ , it could be represented as a multiplication of two matrices

$$A = QR, \quad (4)$$

where  $Q \in \mathbb{R}^{n \times n}$  is an orthonormal matrix and  $R \in \mathbb{R}^{n \times n}$  is an upper-triangular matrix. For the matrix  $A$ , describe how Gram-Schmidt process could be used to find the  $Q$  and  $R$  matrices, and apply this to

$$A = \begin{bmatrix} 3 & -3 & 1 \\ 4 & -4 & -7 \\ 0 & 3 & 3 \end{bmatrix} \quad (5)$$

to find an orthonormal matrix  $Q$  and an upper-triangular matrix  $R$ .

- (b) Given an invertible matrix  $A \in \mathbb{R}^{n \times n}$  and an observation vector  $\vec{b} \in \mathbb{R}^n$ , the solution to the equality

$$A\vec{x} = \vec{b} \quad (6)$$

is given as  $\vec{x} = A^{-1}\vec{b}$ . For the matrix  $A = QR$  from part 4(a), assume that we want to solve

$$A\vec{x} = \begin{bmatrix} 8 \\ -6 \\ 3 \end{bmatrix}. \quad (7)$$

By using the fact that  $Q$  is an orthonormal matrix, find  $\vec{v}$  such that

$$R\vec{x} = \vec{v}. \quad (8)$$

Then, given the upper-triangular matrix  $R$  in part 4(a) and  $\vec{v}$ , find the elements of  $\vec{x}$  sequentially.

- (c) Given an invertible matrix  $B \in \mathbb{R}^{n \times n}$  and an observation vector  $\vec{c} \in \mathbb{R}^n$ , find the computational cost of finding the solution  $\vec{z}$  to the equation  $B\vec{z} = \vec{c}$  by using the  $QR$  decomposition of  $B$ . Assume that  $Q$  and  $R$  matrices are available, and adding, multiplying, and dividing scalars take one unit of “computation”.

As an example, computing the inner product  $\vec{a}^\top \vec{b}$  is said to be  $\mathcal{O}(n)$ , since we have  $n$  scalar multiplication for each  $a_i b_i$ . Similarly, matrix vector multiplication is  $\mathcal{O}(n^2)$ , since matrix vector multiplication can be viewed as computing  $n$  inner products. The computational cost for inverting a matrix in  $\mathbb{R}^n$  is  $\mathcal{O}(n^3)$ , and consequently, the cost grows rapidly as the set of equations grows in size. This is why the expression  $A^{-1}\vec{b}$  is usually not computed by directly inverting the matrix  $A$ . Instead, the  $QR$  decomposition of  $A$  is exploited to decrease the computational cost.

### 5. Determinants

Consider a unit box  $\mathcal{B}$  in  $\mathbb{R}^2$  — i.e., the square with corners  $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$ ,  $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$ ,  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ , and  $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ . Define  $A(\mathcal{B})$  as the parallelogram generated by applying matrix  $A$  to every point in  $\mathcal{B}$ .

- (a) For  $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$ , calculate the location of each corner of  $A(\mathcal{B})$ .

- (b) Write the area of  $A(\mathcal{B})$  as a function of  $\det(A)$ .

*HINT: How are the basis vectors  $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$  and  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$  transformed by the matrix multiplication?*

- (c) Calculate the area of  $A(\mathcal{B})$  for each of the following values of  $A$ .

i.  $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$

ii.  $A = \begin{bmatrix} 2 & 1 \\ 4 & 3 \end{bmatrix}$

iii.  $A = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$

iv.  $A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$

**6. Homework Process**

With whom did you work on this homework? List the names and SIDs of your group members.

*NOTE:* If you didn't work with anyone, you can put "none" as your answer.