

# The Projection Matrix

# Introduction to Projection

- Let  $A$  be an  $n \times d$  matrix.
- In over-determined systems where  $d < n$ , the system  $A\mathbf{x} = \mathbf{b}$  is often inconsistent.
- This occurs when the vector  $\mathbf{b}$  does not lie in the span of the columns of  $A$ .

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- The mapping from  $\mathbf{b}$  to  $\mathbf{b}'$  is a linear transformation defined by the **Projection Matrix**.

# The Least-Squares Objective

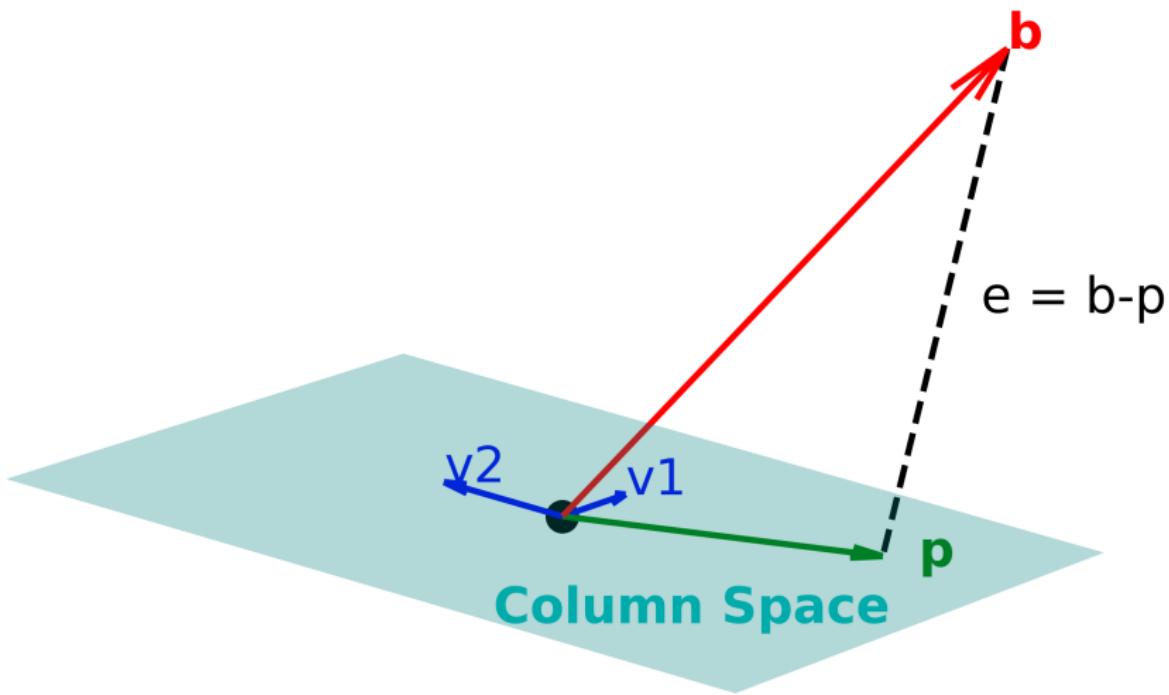
Consider a general  $n \times d$  matrix  $A$  of full rank.

- We seek the coordinate vector  $\mathbf{x}$  such that the projection  $\mathbf{b}' = A\mathbf{x}$  minimizes the squared error:

$$\min_{\mathbf{x}} \|\mathbf{b} - A\mathbf{x}\|^2$$

- The vector  $\mathbf{x}$  represents the coordinates of the projection  $\mathbf{b}'$  in the basis defined by the columns of  $A$ .

## LS: geometric interpretation



# Proof: Derivation of Coordinate Vector $\mathbf{x}$

**Geometric Condition:** The error vector  $(\mathbf{b} - A\mathbf{x})$  represents the shortest path from  $\mathbf{b}$  to the hyperplane. Therefore, it must be **orthogonal** to the hyperplane defined by the columns of  $A$ .

- ① **Orthogonality condition:** The dot product of the error vector with any column of  $A$  must be zero.

$$A^T(\mathbf{b} - A\mathbf{x}) = \mathbf{0}$$

- ② **Expand the terms:**

$$A^T\mathbf{b} - A^TA\mathbf{x} = \mathbf{0}$$

- ③ **Rearrange to form the Normal Equation:**

$$A^TA\mathbf{x} = A^T\mathbf{b}$$

- ④ **Solve for  $\mathbf{x}$**  (assuming columns of  $A$  are linearly independent,  $A^TA$  is invertible):

$$\mathbf{x} = (A^TA)^{-1}A^T\mathbf{b}$$

# The General Projection Matrix

Using the derived expression for  $\mathbf{x}$ , we can find the projection  $\mathbf{b}'$ :

$$\mathbf{b}' = A\mathbf{x} = A \left[ (A^T A)^{-1} A^T \mathbf{b} \right]$$

Grouping the matrix terms gives us the definition of the projection matrix  $P$ :

$$\mathbf{b}' = \underbrace{A(A^T A)^{-1} A^T}_{P} \mathbf{b}$$

## Definition (General Case)

The  $n \times n$  projection matrix  $P$  is defined as:

$$P = A(A^T A)^{-1} A^T$$

## Case: Orthonormal Columns

Consider the case where the columns of the matrix are orthonormal. We denote this matrix as  $Q$ .

- Because the columns are orthonormal, we have  $Q^T Q = I_d$  (Identity matrix of size  $d$ ).
- Substituting  $Q$  into the general formula:

$$P = Q(Q^T Q)^{-1}Q^T = Q(I)^{-1}Q^T = QQ^T$$

### Definition (Orthonormal Case)

$$P = QQ^T$$

Note: While  $Q^T Q = I$ , it is generally true that  $QQ^T \neq I$  for rectangular matrices where  $d < n$ .

# Idempotent Property

A fundamental property of projection is that projecting a vector already on the plane should not change it. Thus, applying  $P$  twice is the same as applying it once ( $P^2 = P$ ).

## Proof (Orthonormal Case)

$$P^2 = (QQ^T)(QQ^T)$$

$$P^2 = Q(Q^T Q)Q^T$$

Since  $Q^T Q = I$ :

$$P^2 = QIQ^T = QQ^T = P$$

**General Property:** Any symmetric matrix satisfying  $P^2 = P$  is a projection matrix.

# Using QR Decomposition

Calculating  $(A^T A)^{-1}$  can be computationally intensive. We can use **QR Decomposition** ( $A = QR$ ) to simplify the computation.

- Multiplying  $A$  by any non-singular matrix does not change the column space, so  $A$  and  $Q$  share the same projection matrix.
- Therefore, we can compute  $P$  using the orthonormal factor  $Q$ :

$$P = QQ^T$$

- The projection is computed as  $\mathbf{b}' = QQ^T \mathbf{b}$ .

## Solving for $\mathbf{x}$

Since  $\mathbf{b}' = QR\mathbf{x}$ , we solve the triangular system:

$$R\mathbf{x} = Q^T \mathbf{b}$$

# Orthogonal Complementary Projections

## Property

If  $P = QQ^T$  is a projection matrix projecting onto a subspace, then the matrix:

$$(I - P)$$

is also a projection matrix.

- It projects onto the **orthogonal complementary vector space** of the column space of  $A$ .
- This captures the residual or error component of the vector  $\mathbf{b}$ .

# Summary

- ① **Projection Matrix:** Linearly transforms  $\mathbf{b}$  to its closest point in the column space of  $A$ .
- ② **Formula:**  $P = A(A^T A)^{-1}A^T$ .
- ③ **Coordinate Vector:**  $\mathbf{x} = (A^T A)^{-1}A^T \mathbf{b}$ .
- ④ **Orthonormal Case:** Simplifies to  $P = QQ^T$ .
- ⑤ **Key Properties:** Symmetric ( $P^T = P$ ) and Idempotent ( $P^2 = P$ ).