

# 7.7. The projection theorem and its implications

Orthogonal projections, formula

# Orthogonal projections to a line in $\mathbb{R}^2$

- Let us obtain a formula for projection to a line containing a nonzero vector  $a$ .
- $x = x_1 + x_2$ ,  $x_1 = ka$ .  $x_2$  is orthogonal to  $a$ .
- $(x - ka) \cdot a = 0$ .  $x \cdot a - k(a \cdot a) = 0$ .  $k = x \cdot a / \|a\|^2$ .
- $x_1 = (x \cdot a / \|a\|^2)a$ .
- $\text{Proj}_a(x) = (x \cdot a / \|a\|^2)a$ .
- Example 1: The matrix expression.

# Orthogonal projections onto a line through 0 in $R^n$ .

**Theorem 7.7.1** *If  $\mathbf{a}$  is a nonzero vector in  $R^n$ , then every vector  $\mathbf{x}$  in  $R^n$  can be expressed in exactly one way as*

$$\mathbf{x} = \mathbf{x}_1 + \mathbf{x}_2 \quad (6)$$

where  $\mathbf{x}_1$  is a scalar multiple of  $\mathbf{a}$  and  $\mathbf{x}_2$  is orthogonal to  $\mathbf{a}$  (and hence to  $\mathbf{x}_1$ ). The vectors  $\mathbf{x}_1$  and  $\mathbf{x}_2$  are given by the formulas

$$\mathbf{x}_1 = \frac{\mathbf{x} \cdot \mathbf{a}}{\|\mathbf{a}\|^2} \mathbf{a} \quad \text{and} \quad \mathbf{x}_2 = \mathbf{x} - \mathbf{x}_1 = \mathbf{x} - \frac{\mathbf{x} \cdot \mathbf{a}}{\|\mathbf{a}\|^2} \mathbf{a} \quad (7)$$

- Proof: Omit

**Definition 7.7.2** *If  $\mathbf{a}$  is a nonzero vector in  $R^n$ , and if  $\mathbf{x}$  is any vector in  $R^n$ , then the **orthogonal projection of  $\mathbf{x}$  onto  $\text{span}\{\mathbf{a}\}$**  is denoted by  $\text{proj}_{\mathbf{a}}\mathbf{x}$  and is defined as*

$$\text{proj}_{\mathbf{a}}\mathbf{x} = \frac{\mathbf{x} \cdot \mathbf{a}}{\|\mathbf{a}\|^2} \mathbf{a} \quad (11)$$

The vector  $\text{proj}_{\mathbf{a}}\mathbf{x}$  is also called the **vector component of  $\mathbf{x}$  along  $\mathbf{a}$** , and  $\mathbf{x} - \text{proj}_{\mathbf{a}}\mathbf{x}$  is called the **vector component of  $\mathbf{x}$  orthogonal to  $\mathbf{a}$** .

# Projection operator in $\mathbb{R}^n$

- $T(\mathbf{x}) = \text{proj}_{\mathbf{a}}(\mathbf{x}) = (\mathbf{x} \cdot \mathbf{a} / \|\mathbf{a}\|^2) \mathbf{a}$
- Orthogonal projection of  $\mathbb{R}^n$  onto  $\text{span}\{\mathbf{a}\}$ .

**Theorem 7.7.3** *If  $\mathbf{a}$  is a nonzero vector in  $\mathbb{R}^n$ , and if  $\mathbf{a}$  is expressed in column form, then the standard matrix for the linear operator  $T(\mathbf{x}) = \text{proj}_{\mathbf{a}} \mathbf{x}$  is*

$$P = \frac{1}{\mathbf{a}^T \mathbf{a}} \mathbf{a} \mathbf{a}^T \quad (16)$$

*This matrix is symmetric and has rank 1.*

- Proof:  $T(\mathbf{e}_j) = ((\mathbf{e}_j \cdot \mathbf{a}) / \|\mathbf{a}\|^2) \mathbf{a} = (a_j / \|\mathbf{a}\|^2) \mathbf{a}$ .
- $P = [a_1 \mathbf{a} \mid a_2 \mathbf{a} \mid \dots \mid a_n \mathbf{a}] / \|\mathbf{a}\|^2 = \mathbf{a} [a_1, a_2, \dots, a_n] / \|\mathbf{a}\|^2 =$
- $\mathbf{a} \mathbf{a}^T / \mathbf{a}^T \mathbf{a}$ .

- If  $a$  is a unit vector  $u$ . Then  $P=uu^T$ .
- Example 4.  $P_\theta$  again
- Example 5.

# Projection theorem

**Theorem 7.7.4 (Projection Theorem for Subspaces)** *If  $W$  is a subspace of  $R^n$ , then every vector  $\mathbf{x}$  in  $R^n$  can be expressed in exactly one way as*

$$\mathbf{x} = \mathbf{x}_1 + \mathbf{x}_2 \quad (20)$$

where  $\mathbf{x}_1$  is in  $W$  and  $\mathbf{x}_2$  is in  $W^\perp$ .

- Proof: Let  $\{w_1, \dots, w_k\}$  be the basis of  $W$ . Let  $M$  be the  $n \times k$  matrix with columns  $w_1, w_2, \dots, w_k$ .  $k \leq n$ .
  - $W$  column space of  $M$ .  $W^\perp$  null space of  $M^T$ .
  - Write  $x = x_1 + x_2$ ,  $x_1$  in  $W$  and  $x_2$  in  $W^\perp$ .
  - $x_1 = Mv$  and  $M^T(x_2) = 0$  or  $M^T(x - x_2) = 0$ .
  - $M^T(x - Mv) = 0$ .
  - This has a unique solution  $\leftrightarrow x_1, x_2$  exist and are unique.

- Rewrite  $M^T M v = M^T x$ .
- $M^T M$  is  $k \times k$ -matrix
- $M$  has a full column rank as  $w_1, \dots, w_k$  are independent.
- $M^T M$  is invertible by Theorem 7.5.10.
- $v = (M^T M)^{-1} M^T x$ .
- $x = \text{proj}_W(x) + \text{proj}_{W^c}(x)$ .
- Since  $\text{proj}_W(x) = x_1 = Mv$ , we have

**Theorem 7.7.5** *If  $W$  is a nonzero subspace of  $R^n$ , and if  $M$  is any matrix whose column vectors form a basis for  $W$ , then*

$$\text{proj}_W \mathbf{x} = M(M^T M)^{-1} M^T \mathbf{x} \quad (25)$$

*for every column vector  $\mathbf{x}$  in  $R^n$ .*

- $T(x) = \text{Proj}_W(x) = M(M^T M)^{-1} M^T(x)$ .
- Matrix is  $P = M(M^T M)^{-1} M^T$
- Orthogonal projection of  $\mathbb{R}^n$  to  $W$ .
- This extends the previous formula.
- Example 6.

# Condition for orthogonal projection

- $P^T = (M(M^T M)^{-1} M^T)^T = M((M^T M)^{-1})^T M^T = M(M^T M)^{-1} M^T = P.$
- $P^2 = M(M^T M)^{-1} M^T (M(M^T M)^{-1} M^T) = M(M^T M)^{-1} (M^T M) (M^T M)^{-1} M^T = M(M^T M)^{-1} M^T = P.$
- $P^2 = P.$

**Theorem 7.7.6** *An  $n \times n$  matrix  $P$  is the standard matrix for an orthogonal projection of  $R^n$  onto a  $k$ -dimensional subspace of  $R^n$  if and only if  $P$  is symmetric, idempotent, and has rank  $k$ . The subspace  $W$  is the column space of  $P$ .*

# Strang diagrams

- $Ax=b$ . A  $m \times n$  matrix
- $\text{row}(A)$ ,  $\text{null}(A)$  are orthogonal complements
- $\text{col}(A)$  and  $\text{null}(A^T)$  are orthogonal complements.
- $x = x_{\text{row}(A)} + x_{\text{null}(A)}$ .
- $b = b_{\text{col}(A)} + b_{\text{null}(A^T)}$ .
- $Ax=b$  is consistent iff  $b_{\text{null}(A^T)}=0$ .
- See Fig. 7.7.6.

**Theorem 7.7.7** Suppose that  $A$  is an  $m \times n$  matrix and  $\mathbf{b}$  is in the column space of  $A$ .

- (a) If  $A$  has full column rank, then the system  $A\mathbf{x} = \mathbf{b}$  has a unique solution, and that solution is in the row space of  $A$ .
- (b) If  $A$  does not have full column rank, then the system  $A\mathbf{x} = \mathbf{b}$  has infinitely many solutions, but there is a unique solution in the row space of  $A$ . Moreover, among all solutions of the system, the solution in the row space of  $A$  has the smallest norm.

- Proof: (a) If  $A$  has a full rank, Theorem 7.5.6 implies  $Ax=b$  has a unique solution or is inconsistent. Since  $b$  is in  $\text{col}(A)$ , it is consistent.
- (b). Theorem 7.5.6 implies  $Ax=0$  has infinitely many solutions. Smallest norm  $\rightarrow$  omit.

# Double perp theorem.

**Theorem 7.7.8 (Double Perp Theorem)** *If  $W$  is a subspace of  $R^n$ , then  $(W^\perp)^\perp = W$ .*

- Proof: Show  $W$  is a subset of  $(W^c)^c$ :
  - Suppose  $w$  is in  $W$ . Then  $w$  is perp to every  $a$  in  $W^c$ . This means that  $w$  is in  $(W^c)^c$ .
- Show  $(W^c)^c$  is a subset of  $W$ .
  - Let  $w$  be in  $(W^c)^c$ .
  - Write  $w=w_1+w_2$ ,  $w_1$  in  $W$ ,  $w_2$  in  $W^c$ .
  - $w_2 \cdot w=0$ .
  - $(w_2 \cdot w_1)+(w_2 \cdot w_2)=0$ .  $w_2 \cdot w_2=0 \rightarrow w_2 =0$ .

# Orthogonal projection to $W^c$

- $\text{Proj}_{W^c}(x) = x - \text{proj}_W(x) = Ix - Px = (I - P)x.$
- Thus the matrix of  $\text{Proj}_{W^c}$  is  $I - P = I - M(M^T M)^{-1} M^T.$
- $I - P$  is also symmetric and idempotent.
- $\text{Rank}(I - P) = n - \text{rank } P.$