MATH 304 Linear Algebra

Lecture 29: Orthogonal projection (continued). Least squares problems.

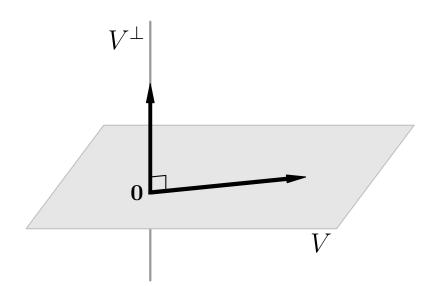
## **Orthogonal complement**

*Definition.* Let  $S \subset \mathbb{R}^n$ . The **orthogonal complement** of *S*, denoted  $S^{\perp}$ , is the set of all vectors  $\mathbf{x} \in \mathbb{R}^n$  that are orthogonal to *S*.

**Theorem 1 (i)**  $S^{\perp}$  is a subspace of  $\mathbb{R}^n$ . (ii)  $(S^{\perp})^{\perp} = \operatorname{Span}(S)$ .

**Theorem 2** If V is a subspace of  $\mathbb{R}^n$ , then (i)  $(V^{\perp})^{\perp} = V$ , (ii)  $V \cap V^{\perp} = \{\mathbf{0}\}$ , (iii) dim  $V + \dim V^{\perp} = n$ .

**Theorem 3** If V is the row space of a matrix, then  $V^{\perp}$  is the nullspace of the same matrix.



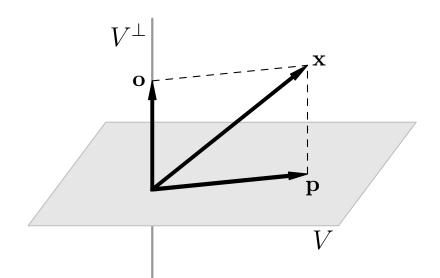
## **Orthogonal projection**

**Theorem 1** Let V be a subspace of  $\mathbb{R}^n$ . Then any vector  $\mathbf{x} \in \mathbb{R}^n$  is uniquely represented as  $\mathbf{x} = \mathbf{p} + \mathbf{o}$ , where  $\mathbf{p} \in V$  and  $\mathbf{o} \in V^{\perp}$ .

In the above expansion,  $\mathbf{p}$  is called the **orthogonal projection** of the vector  $\mathbf{x}$  onto the subspace V.

**Theorem 2**  $\|\mathbf{x} - \mathbf{v}\| > \|\mathbf{x} - \mathbf{p}\|$  for any  $\mathbf{v} \neq \mathbf{p}$  in V.

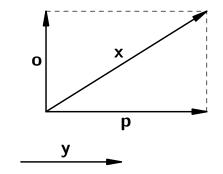
Thus  $\|\mathbf{o}\| = \|\mathbf{x} - \mathbf{p}\| = \min_{\mathbf{v} \in V} \|\mathbf{x} - \mathbf{v}\|$  is the **distance** from the vector  $\mathbf{x}$  to the subspace V.



### Orthogonal projection onto a vector

Let 
$$\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$$
, with  $\mathbf{y} \neq \mathbf{0}$ .

Then there exists a unique decomposition  $\mathbf{x} = \mathbf{p} + \mathbf{o}$ such that  $\mathbf{p}$  is parallel to  $\mathbf{y}$  and  $\mathbf{o}$  is orthogonal to  $\mathbf{y}$ .



 $\mathbf{p} = \text{orthogonal projection of } \mathbf{x} \text{ onto } \mathbf{y}$ 

#### Orthogonal projection onto a vector

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We have  $\mathbf{p} = \alpha \mathbf{y}$  for some  $\alpha \in \mathbb{R}$ . Then

$$\mathbf{0} = \mathbf{o} \cdot \mathbf{y} = (\mathbf{x} - \alpha \mathbf{y}) \cdot \mathbf{y} = \mathbf{x} \cdot \mathbf{y} - \alpha \mathbf{y} \cdot \mathbf{y}.$$

$$\implies \alpha = \frac{\mathbf{x} \cdot \mathbf{y}}{\mathbf{y} \cdot \mathbf{y}} \implies \mathbf{p} = \frac{\mathbf{x} \cdot \mathbf{y}}{\mathbf{y} \cdot \mathbf{y}} \mathbf{y}$$

**Problem.** Find the distance from the point  $\mathbf{x} = (3, 1)$  to the line spanned by  $\mathbf{y} = (2, -1)$ .

Consider the decomposition  $\mathbf{x} = \mathbf{p} + \mathbf{o}$ , where  $\mathbf{p}$  is parallel to  $\mathbf{y}$  while  $\mathbf{o} \perp \mathbf{y}$ . The required distance is the length of the orthogonal component  $\mathbf{o}$ .

$$\mathbf{p} = \frac{\mathbf{x} \cdot \mathbf{y}}{\mathbf{y} \cdot \mathbf{y}} \mathbf{y} = \frac{5}{5} (2, -1) = (2, -1),$$
  
$$\mathbf{o} = \mathbf{x} - \mathbf{p} = (3, 1) - (2, -1) = (1, 2), \quad ||\mathbf{o}|| = \sqrt{5}.$$

**Problem.** Find the point on the line y = -x that is closest to the point (3, 4).

The required point is the projection **p** of  $\mathbf{v} = (3, 4)$  on the vector  $\mathbf{w} = (1, -1)$  spanning the line y = -x.

$$\mathbf{p} = rac{\mathbf{v} \cdot \mathbf{w}}{\mathbf{w} \cdot \mathbf{w}} \mathbf{w} = rac{-1}{2} (1, -1) = \left(-rac{1}{2}, rac{1}{2}
ight).$$

**Problem.** Let  $\Pi$  be the plane spanned by vectors  $\mathbf{v}_1 = (1, 1, 0)$  and  $\mathbf{v}_2 = (0, 1, 1)$ . (i) Find the orthogonal projection of the vector  $\mathbf{x} = (4, 0, -1)$  onto the plane  $\Pi$ . (ii) Find the distance from  $\mathbf{x}$  to  $\Pi$ .

We have  $\mathbf{x} = \mathbf{p} + \mathbf{o}$ , where  $\mathbf{p} \in \Pi$  and  $\mathbf{o} \perp \Pi$ . Then the orthogonal projection of  $\mathbf{x}$  onto  $\Pi$  is  $\mathbf{p}$  and the distance from  $\mathbf{x}$  to  $\Pi$  is  $\|\mathbf{o}\|$ .

We have  $\mathbf{p} = \alpha \mathbf{v}_1 + \beta \mathbf{v}_2$  for some  $\alpha, \beta \in \mathbb{R}$ . Then  $\mathbf{o} = \mathbf{x} - \mathbf{p} = \mathbf{x} - \alpha \mathbf{v}_1 - \beta \mathbf{v}_2$ .

$$\begin{cases} \mathbf{o} \cdot \mathbf{v}_1 = \mathbf{0} \\ \mathbf{o} \cdot \mathbf{v}_2 = \mathbf{0} \end{cases} \iff \begin{cases} \alpha(\mathbf{v}_1 \cdot \mathbf{v}_1) + \beta(\mathbf{v}_2 \cdot \mathbf{v}_1) = \mathbf{x} \cdot \mathbf{v}_1 \\ \alpha(\mathbf{v}_1 \cdot \mathbf{v}_2) + \beta(\mathbf{v}_2 \cdot \mathbf{v}_2) = \mathbf{x} \cdot \mathbf{v}_2 \end{cases}$$

$$\mathbf{x}=(4,0,-1)$$
,  $\mathbf{v}_1=(1,1,0)$ ,  $\mathbf{v}_2=(0,1,1)$ 

$$\begin{cases} \alpha(\mathbf{v}_1 \cdot \mathbf{v}_1) + \beta(\mathbf{v}_2 \cdot \mathbf{v}_1) = \mathbf{x} \cdot \mathbf{v}_1 \\ \alpha(\mathbf{v}_1 \cdot \mathbf{v}_2) + \beta(\mathbf{v}_2 \cdot \mathbf{v}_2) = \mathbf{x} \cdot \mathbf{v}_2 \end{cases}$$
$$\iff \begin{cases} 2\alpha + \beta = 4 \\ \alpha + 2\beta = -1 \end{cases} \iff \begin{cases} \alpha = 3 \\ \beta = -2 \end{cases}$$

$$\mathbf{p} = 3\mathbf{v}_1 - 2\mathbf{v}_2 = (3, 1, -2)$$
  
 $\mathbf{o} = \mathbf{x} - \mathbf{p} = (1, -1, 1)$   
 $\|\mathbf{o}\| = \sqrt{3}$ 

**Problem.** Let  $\Pi$  be the plane spanned by vectors  $\mathbf{v}_1 = (1, 1, 0)$  and  $\mathbf{v}_2 = (0, 1, 1)$ . (i) Find the orthogonal projection of the vector  $\mathbf{x} = (4, 0, -1)$  onto the plane  $\Pi$ . (ii) Find the distance from  $\mathbf{x}$  to  $\Pi$ .

Alternative solution: We have  $\mathbf{x} = \mathbf{p} + \mathbf{o}$ , where  $\mathbf{p} \in \Pi$  and  $\mathbf{o} \perp \Pi$ . Then the orthogonal projection of  $\mathbf{x}$  onto  $\Pi$  is  $\mathbf{p}$  and the distance from  $\mathbf{x}$  to  $\Pi$  is  $\|\mathbf{o}\|$ .

Notice that **o** is the orthogonal projection of **x** onto the orthogonal complement  $\Pi^{\perp}$ . In the previous lecture, we found that  $\Pi^{\perp}$  is the line spanned by the vector  $\mathbf{y} = (1, -1, 1)$ . It follows that

$$\mathbf{o} = \frac{\mathbf{x} \cdot \mathbf{y}}{\mathbf{y} \cdot \mathbf{y}} \mathbf{y} = \frac{3}{3} (1, -1, 1) = (1, -1, 1).$$

Then  $\mathbf{p} = \mathbf{x} - \mathbf{o} = (4, 0, -1) - (1, -1, 1) = (3, 1, -2)$  and  $\|\mathbf{o}\| = \sqrt{3}$ .

Overdetermined system of linear equations:

$$\begin{cases} x + 2y = 3 \\ 3x + 2y = 5 \\ x + y = 2.09 \end{cases} \iff \begin{cases} x + 2y = 3 \\ -4y = -4 \\ -y = -0.91 \end{cases}$$

No solution: inconsistent system

Assume that a solution  $(x_0, y_0)$  does exist but the system is not quite accurate, namely, there may be some errors in the right-hand sides.

**Problem.** Find a good approximation of  $(x_0, y_0)$ .

One approach is the **least squares fit**. Namely, we look for a pair (x, y) that minimizes the sum  $(x+2y-3)^2 + (3x+2y-5)^2 + (x+y-2.09)^2$ .

# Least squares solution

System of linear equations:  

$$\begin{cases}
a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\
a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\
\dots \dots \dots \\
a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m
\end{cases} \iff A\mathbf{x} = \mathbf{b}$$
For any  $\mathbf{x} \in \mathbb{R}^n$  define a **residual**  $r(\mathbf{x}) = \mathbf{b} - A\mathbf{x}$ .  
The **least squares solution x** to the system is the one that minimizes  $||r(\mathbf{x})||$  (or, equivalently,  $||r(\mathbf{x})||^2$ ).

$$\|r(\mathbf{x})\|^2 = \sum_{i=1}^m (a_{i1}x_1 + a_{i2}x_2 + \cdots + a_{in}x_n - b_i)^2$$

Let A be an  $m \times n$  matrix and let  $\mathbf{b} \in \mathbb{R}^m$ .

**Theorem** A vector  $\hat{\mathbf{x}}$  is a least squares solution of the system  $A\mathbf{x} = \mathbf{b}$  if and only if it is a solution of the associated **normal system**  $A^T A \mathbf{x} = A^T \mathbf{b}$ .

*Proof:*  $A\mathbf{x}$  is an arbitrary vector in R(A), the column space of A. Hence the length of  $r(\mathbf{x}) = \mathbf{b} - A\mathbf{x}$  is minimal if  $A\mathbf{x}$  is the orthogonal projection of  $\mathbf{b}$  onto R(A). That is, if  $r(\mathbf{x})$  is orthogonal to R(A).

We know that {row space}<sup> $\perp$ </sup> = {nullspace} for any matrix. In particular,  $R(A)^{\perp} = N(A^{T})$ , the nullspace of the transpose matrix of A. Thus  $\hat{\mathbf{x}}$  is a least squares solution if and only if

$$A^T r(\hat{\mathbf{x}}) = \mathbf{0} \iff A^T (\mathbf{b} - A\hat{\mathbf{x}}) = \mathbf{0} \iff A^T A \hat{\mathbf{x}} = A^T \mathbf{b}.$$

**Corollary** The normal system  $A^T A \mathbf{x} = A^T \mathbf{b}$  is always consistent.

Problem. Find the least squares solution to

$$\begin{cases} x + 2y = 3\\ 3x + 2y = 5\\ x + y = 2.09 \end{cases}$$
$$\begin{pmatrix} 1 & 2\\ 3 & 2\\ 1 & 1 \end{pmatrix} \begin{pmatrix} x\\ y \end{pmatrix} = \begin{pmatrix} 3\\ 5\\ 2.09 \end{pmatrix}$$
$$\begin{pmatrix} 1 & 3 & 1\\ 2 & 2 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2\\ 3 & 2\\ 1 & 1 \end{pmatrix} \begin{pmatrix} x\\ y \end{pmatrix} = \begin{pmatrix} 1 & 3 & 1\\ 2 & 2 & 1 \end{pmatrix} \begin{pmatrix} 3\\ 5\\ 2.09 \end{pmatrix}$$
$$\begin{pmatrix} 11 & 9\\ 9 & 9 \end{pmatrix} \begin{pmatrix} x\\ y \end{pmatrix} = \begin{pmatrix} 20.09\\ 18.09 \end{pmatrix} \iff \begin{cases} x = 1\\ y = 1.01 \end{cases}$$