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Math 22 –  
Linear Algebra and its  
applications

- Lecture 21 -

**Instructor:** Bjoern Muetzel

# GENERAL INFORMATION

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- **Office hours:** Tu 1-3 pm, Th, Sun 2-4 pm in KH 229

**Tutorial:** Tu, Th, Sun 7-9 pm in KH 105

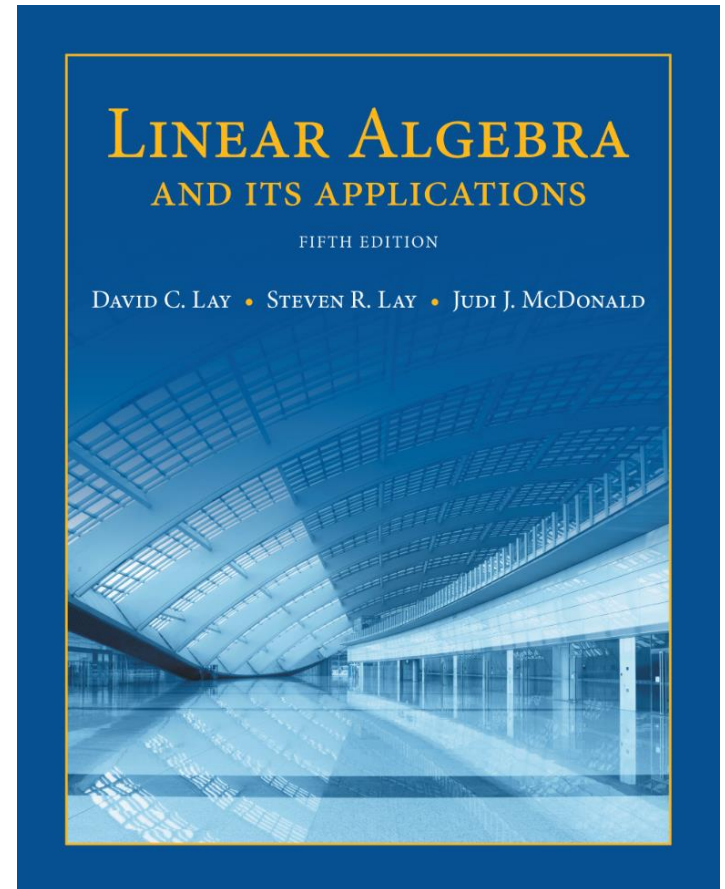
- **Midterm 2:** today at 4 pm in Carpenter 013
  - **Topics: Chapter 2.1 – 4.7** (included)
  - about **8-9** questions
  - **Practice exam 2 solutions** available

# 6

## Orthogonality and Least Squares

### 6.2

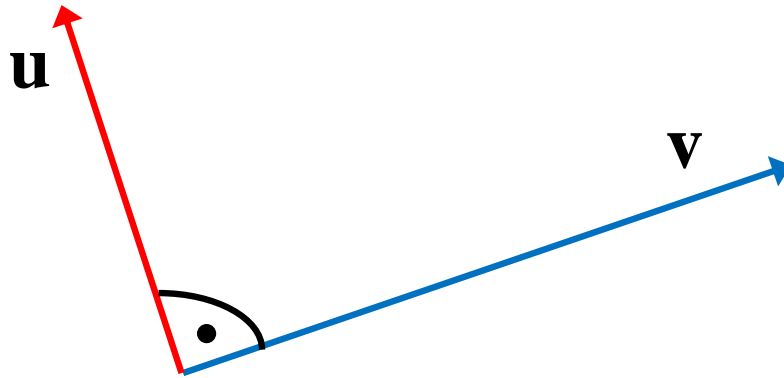
#### ORTHOGONAL SETS



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## Summary:

If the vectors of a set  $\mathbf{S}$  are **orthogonal**, then they are pairwise orthogonal to each other. If these **vectors** are additionally **normalized** then they look like a standard basis.



# GEOMETRIC INTERPRETATION

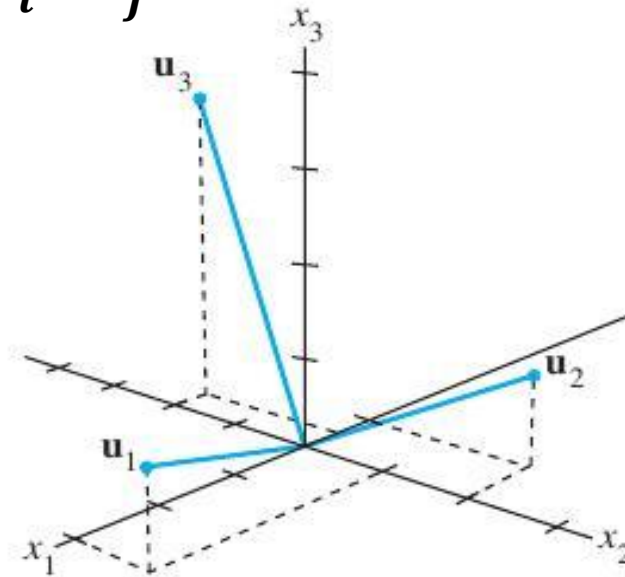
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# ORTHOGONAL SETS

- **Definition:** A set of vectors  $\{u_1, \dots, u_p\}$  in  $\mathbb{R}^n$  is said to be an **orthogonal set** if these vectors are pairwise orthogonal, that is, if

$$\mathbf{u}_i \cdot \mathbf{u}_j = 0 \quad \text{whenever} \quad i \neq j.$$



- **Theorem 4:** If  $S = \{u_1, \dots, u_p\}$  is an orthogonal set of **nonzero** vectors in  $\mathbb{R}^n$ , then  $S$  is linearly independent and hence is a basis for the subspace spanned by  $S$ .

**Proof:**

**Definition:** An **orthogonal basis** for a subspace  $W$  of  $\mathbb{R}^n$  is a basis for  $W$  that is also an orthogonal set.

**Note: Theorem 4** says that an orthogonal set  $S$  of nonzero vectors is automatically a basis for  $\text{Span}\{S\}$ .



# ORTHOGONAL SETS

- **Theorem 5:** Let  $\{u_1, \dots, u_p\}$  be an orthogonal basis for a subspace  $W$  of  $\mathbb{R}^n$ . For each  $y$  in  $W$ , the weights in the linear combination are given by

$$y = x_1 u_1 + \dots + x_p u_p, \quad \text{where} \quad x_j = \frac{y \cdot u_j}{u_j \cdot u_j} \quad \text{for all } j \text{ in } \{1, \dots, p\}.$$

**Note:** For the matrix  $U = [u_1, u_2, \dots, u_p]$  and  $y$  in  $W$  this means that we can immediately write down the **solution  $x$**  for  $\mathbf{U}x = y$ .

**Proof:**

**Example:** Let  $u_1 = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$  and  $u_2 = \begin{bmatrix} -3 \\ 1 \end{bmatrix}$  in  $\mathbb{R}^2$  and  $U = [u_1, u_2]$ .

1.) Show that  $u_1$  and  $u_2$  are orthogonal.

2.) For  $y = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ , find  $x$ , such that  $Ux = y$  using **Theorem 5**.

# ORTHOGONAL PROJECTION

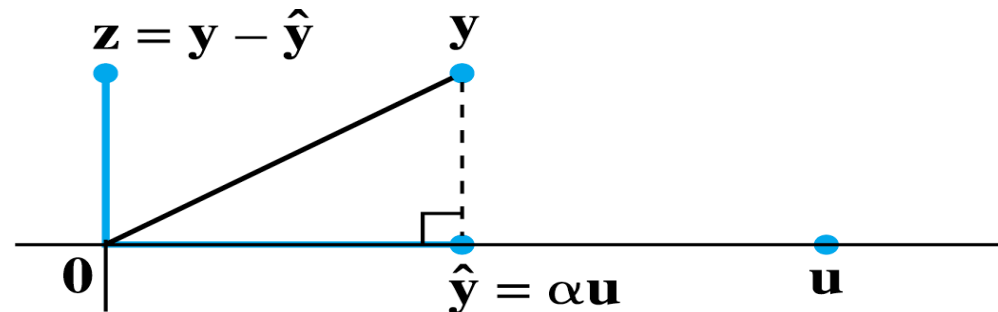
- **Theorem:** Let  $u$  be a nonzero vector in  $\mathbb{R}^n$  and  $L = \text{Span}\{u\}$ . Then the **orthogonal projection** of a vector  $y$  in  $\mathbb{R}^n$  onto  $u$  (or  $L$ ) is

$$\hat{y} = \text{proj}_L(y) = \frac{y \cdot u}{u \cdot u} u.$$

The **component of  $y$  orthogonal to  $u$**  is

$$z = y - \hat{y}.$$

- **Example:** Orthogonal projection in  $\mathbb{R}^2$ .



Finding  $\alpha$  to make  $y - \hat{y}$  orthogonal to  $u$ .

- **Note:** The vectors  $z$  and  $\hat{y}$  are orthogonal as  $\hat{y}$  is in  $\text{Span}\{u\}$  and  $z$  is orthogonal to  $u$ .
- **Proof of the Theorem:**

# ORTHOGONAL PROJECTION

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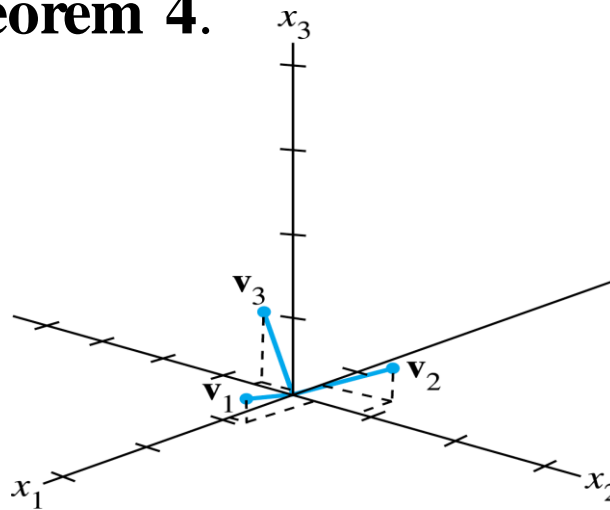
**Example:** Let  $y = \begin{bmatrix} 7 \\ 6 \end{bmatrix}$  and  $u = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$  be vectors in  $\mathbb{R}^2$ .

- 1.) Find the orthogonal projection  $\hat{y}$  of  $y$  onto  $u$ .
- 2.) Write  $y$  as the sum of the two orthogonal vectors  $\hat{y}$  in  $\text{Span}\{u\}$  and  $z$ , which is orthogonal to  $u$ .
- 3.) Draw a picture of the vectors  $y$ ,  $u$ ,  $\hat{y}$  and  $z$  in  $\mathbb{R}^2$ .



# ORTHONORMAL SETS

- **Definition:** A set  $\{\mathbf{u}_1, \dots, \mathbf{u}_p\}$  is an **orthonormal set** if it is an orthogonal set of unit vectors.
- If  $W$  is the subspace spanned by such a set, then  $\{\mathbf{u}_1, \dots, \mathbf{u}_p\}$  is an **orthonormal basis** for  $W$ , since the set is automatically linearly independent, by **Theorem 4**.



- **Note 1:** The simplest example of an orthonormal set is the standard basis  $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$  for  $\mathbb{R}^n$  or subsets of  $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ .

- **Note 2:** When the vectors in an **orthogonal set** of nonzero vectors are **normalized** to have unit length, the **new vectors will be an orthonormal set**.

- **Example:** Let  $v_1 = \frac{1}{\sqrt{11}} \cdot \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix}$ ,  $v_2 = \frac{1}{\sqrt{6}} \cdot \begin{bmatrix} -1 \\ 2 \\ 1 \end{bmatrix}$  and  $v_3 = \frac{1}{\sqrt{66}} \cdot \begin{bmatrix} -1 \\ -4 \\ 7 \end{bmatrix}$ .

Show that  $\{v_1, v_2, v_3\}$  is an orthonormal basis of  $\mathbb{R}^3$ , then draw a picture.



# ORTHONORMAL SETS

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# ORTHONORMAL SETS

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- **Theorem 6:** An  $m \times n$  matrix  $U$ , where  $n \leq m$  has orthonormal columns if and only if  $U^T U = I_n$ .
- **Proof:** Let  $U = [u_1, u_2, \dots, u_n]$  and compute  $U^T U$ .

# ORTHONORMAL SETS

- **Theorem 7:** Let  $U$  be an  $m \times n$  matrix, where  $n \leq m$  with orthonormal columns, and let  $x$  and  $y$  be in  $\mathbb{R}^n$ . Then
  - $(Ux) \cdot (Uy) = x \cdot y$ .
  - $\|Ux\| = \|x\|$ .
  - $(Ux) \cdot (Uy) = 0$  if and only if  $x \cdot y = 0$ .
- **Note:** Properties (a) and (c) say that the linear mapping  $x \mapsto Ux$  preserves **lengths, distance** and **orthogonality**.
- **Proof of Theorem 7:**

