

# TMA 4115 Matematikk 3

Lecture 26 for MBIOT5, MTKJ, MTNANO

Alexander Schmeding

NTNU

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## Inner product, length and orthogonality

Let  $\vec{x}, \vec{y}$  be vectors in  $\mathbb{R}^n$  with components  $x_i$  and  $y_i$ , respectively.

**Dot product/ Inner product**  $\vec{x} \cdot \vec{y} = \vec{x}^T \vec{y} = \sum_{i=1}^n x_i y_i$ .

Define **Orthogonality** of vectors:  $\vec{x}$  and  $\vec{y}$  are orthogonal to each other if and only if  $\vec{x} \cdot \vec{y} = 0$ .

For a subspace  $W \subseteq \mathbb{R}^n$  we can split  $\vec{y} \in \mathbb{R}^n$  into

$$\vec{y} = \vec{y}_W + \vec{z}, \quad \vec{y}_W \in W, \quad \vec{z} \in W^\perp$$

and obtain an orthogonal projection  $\text{proj}_W: \mathbb{R}^n \rightarrow W, \vec{y} \mapsto \vec{y}_W$ .

If  $\{ \vec{v}_1, \dots, \vec{v}_p \}$  is an orthogonal basis for  $W$ , its easy to compute

$$\vec{y}_W = \sum_{i=1}^p \frac{\vec{y} \cdot \vec{v}_i}{\vec{v}_i \cdot \vec{v}_i} \vec{v}_i$$

## 21.7 The Gram-Schmidt Process

Let  $\{\vec{x}_1, \dots, \vec{x}_p\}$  be a basis for a non-zero subspace  $W \subseteq \mathbb{R}^n$ .  
Define

$$\begin{aligned}\vec{v}_1 &= \vec{x}_1 \\ \vec{v}_2 &= \vec{x}_2 - \frac{\vec{x}_2 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 \\ \vec{v}_3 &= \vec{x}_3 - \frac{\vec{x}_3 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 - \frac{\vec{x}_3 \cdot \vec{v}_2}{\vec{v}_2 \cdot \vec{v}_2} \vec{v}_2 \\ &\vdots = \quad \quad \quad \vdots \\ \vec{v}_p &= \vec{x}_p - \sum_{i=1}^{p-1} \frac{\vec{x}_p \cdot \vec{v}_i}{\vec{v}_i \cdot \vec{v}_i} \vec{v}_i\end{aligned}$$

Then  $\{\vec{v}_1, \dots, \vec{v}_p\}$  is an orthogonal basis for  $W$  and in addition

$$\text{span}\{\vec{v}_1, \dots, \vec{v}_k\} = \text{span}\{\vec{x}_1, \dots, \vec{x}_k\} \quad \text{for } 1 \leq k \leq p$$

## Application for Gram-Schmidt

(Spring 2011 Problem 5)

(a) Find an orthogonal basis of  $\text{Col}(A)$  with

$$A = \begin{bmatrix} 1 & 3 & 0 & 1 \\ 2 & 1 & 5 & -3 \\ -1 & -1 & -2 & 1 \end{bmatrix}$$

(b) Find for  $\vec{y} = \begin{bmatrix} 1 \\ 7 \\ 3 \end{bmatrix}$  the nearest point in  $\text{Col}(A)$

To run Gram-Schmidt we need a basis of  $\text{Col}(A)$   
( $\rightarrow$  Gauss elimination on  $A$ !)

Alternatively, we can apply Gram-Schmidt to a generating system of  $\text{Col}(A)$ . This is less work!

Let  $\vec{x}_i$  be the  $i$ -th column of  $A$ .

## Gram Schmidt on a generating system

We note that all vectors in the generating set are non-zero. Thus

$$\text{set } \vec{v}_1 = \vec{x}_1 = \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix} \text{ then } \vec{v}_2 = \vec{x}_2 - \frac{\vec{x}_2 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 = \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix},$$

$$\vec{v}_3 = \vec{x}_3 - \frac{\vec{x}_3 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 - \frac{\vec{x}_3 \cdot \vec{v}_2}{\vec{v}_2 \cdot \vec{v}_2} \vec{v}_2 = \vec{0}$$

we get  $\vec{0}$ , so  $\vec{x}_3$  was already contained in the span of  $\vec{v}_1, \vec{v}_2$   
 $\rightsquigarrow$  discard  $\vec{v}_3$  and continue with the next vector  $\vec{x}_4$ .

$$\vec{v}_3 = \vec{x}_4 - \frac{\vec{x}_4 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 - \frac{\vec{x}_4 \cdot \vec{v}_2}{\vec{v}_2 \cdot \vec{v}_2} \vec{v}_2 = \vec{0}$$

again the result is  $\vec{0}$  and we discard  $\vec{x}_4$ . Since it was the last, the

orthogonal basis of  $\text{Col}(A)$  is  $\left\{ \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix}, \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix} \right\}$

## Application of Gram-Schmidt

To answer part (b), we recall the

**21.6 Best approximation theorem** Let  $W \subseteq \mathbb{R}^n$  be a subspace and  $\vec{y} \in \mathbb{R}^n$ . Then  $\vec{y}_W = \text{proj}_W(\vec{y})$  is closest in  $W$  to  $\vec{y}$ :

$$\|\vec{y} - \vec{y}_W\| < \|\vec{y} - \vec{v}\|, \quad \vec{v} \in W \setminus \{\vec{y}_W\}$$

We need to compute  $\vec{y}_{\text{Col}(A)}$  as the nearest point to  $\vec{y}$  in  $\text{Col}(A)$ .

$$\vec{y}_{\text{Col}(A)} = \frac{\vec{y} \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 + \frac{\vec{y} \cdot \vec{v}_2}{\vec{v}_2 \cdot \vec{v}_2} \vec{v}_2 = \begin{bmatrix} 0 \\ 5 \\ -2 \end{bmatrix}$$

# Weight of baggage

We want to know the weight  $x$  of our suitcase.

If it is too heavy we have to pay extra on the airplane and we don't want that.



On a scale



weight  $x = x_1 \text{ kg}$

Second scale



weight  $x = x_2 \text{ kg}$

Third scale



weight  $x = x_3 \text{ kg}$

What is the weight of our suitcase?

## Weight of baggage II

From our scales we have the information:

$$\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \leftrightarrow \begin{cases} x = x_1 \\ x = x_2 \\ x = x_3 \end{cases}$$

Unless  $x_1 = x_2 = x_3$  holds (by chance), the linear system is **inconsistent**, i.e. there is no exact solution.

**Problem:** We want an “approximate” solution. This means a number  $y$  such that the errors  $|x_i - y|$  are as small as possible.



## Weight of baggage III

**Idea:** If  $A\vec{x} = \vec{b}$  is inconsistent, approximate solution means an  $\vec{x}$  such that  $A\vec{x}$  is as near as possible to the target vector  $\vec{b}$ .

$A\vec{x}$  minimizes the distance to  $\vec{b}$  if it is  $\text{proj}_{\text{Col}(A)}(\vec{b})$ !

This is in the column space, and we can then solve

$$A\vec{x} = \text{proj}_{\text{Col}(A)}(\vec{b}).$$

In our example with  $A = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$  and  $\vec{b} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$ , we set  $\vec{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$

and obtain

$$\text{proj}_{\text{Col}(A)}(\vec{b}) = \frac{\vec{b} \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 = \begin{bmatrix} \frac{x_1+x_2+x_3}{3} \\ \frac{x_1+x_2+x_3}{3} \\ \frac{x_1+x_2+x_3}{3} \end{bmatrix}$$

Thus  $Ax = \text{proj}_{\text{Col}(A)}(\vec{b})$  shows:  $x = \frac{x_1+x_2+x_3}{3}$ .