

TMA 4115 Matematikk 3

Lecture 25 for MTFYMA

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13. April 2016

In today's lecture we will ...

- learn more about orthogonality and orthogonal complements
- study orthogonal projection
- use the Gram-Schmidt algorithm to construct orthogonal (/orthonormal) bases

Inner product, length and orthogonality

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

Dot product/ Inner product

$$\vec{x} \cdot \vec{y} = \vec{x}^T \vec{y} = \sum_{i=1}^n x_i y_i.$$

Can use this to define

- **Length** of a vector: $\|\vec{x}\| = \sqrt{\vec{x} \cdot \vec{x}}$,
- **Distance** between \vec{u} and \vec{v} : $\text{dist}(\vec{u}, \vec{v}) = \|\vec{u} - \vec{v}\|$,
- **Orthogonality** of vectors: \vec{x} and \vec{y} are orthogonal to each other if and only if $\vec{x} \cdot \vec{y} = 0$.

Orthogonal Complement

Let $W \subseteq \mathbb{R}^n$ be non-empty. We say

- $\vec{z} \in \mathbb{R}^n$ is orthogonal to W if for all $\vec{v} \in W$, $\vec{v} \cdot \vec{z} = 0$
- W^\perp is the set of all vectors in \mathbb{R}^n orthogonal to W
(**orthogonal complement** of W).

Example:

For $\vec{0} \in \mathbb{R}^n$ we have $\{\vec{0}\}^\perp = \mathbb{R}^n$ and $(\mathbb{R}^n)^\perp = \{\vec{0}\}$.

$$L = \text{span} \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\} \subseteq \mathbb{R}^2 \text{ then } L^\perp = \text{span} \left\{ \begin{bmatrix} -1 \\ 1 \end{bmatrix} \right\}.$$

$$\text{furthermore } \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}^\perp = \text{span} \left\{ \begin{bmatrix} -1 \\ 1 \end{bmatrix} \right\} = L^\perp$$

$$\text{If } P \text{ is the } x - y\text{-plane in } \mathbb{R}^3 \text{ then } P^\perp = \text{span} \left\{ \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$$

Orthogonal complement

20.14 Theorem

$W \subseteq \mathbb{R}^n$ a subspace with $W = \text{span}\{u_1, \dots, u_p\}$. Then the following holds:

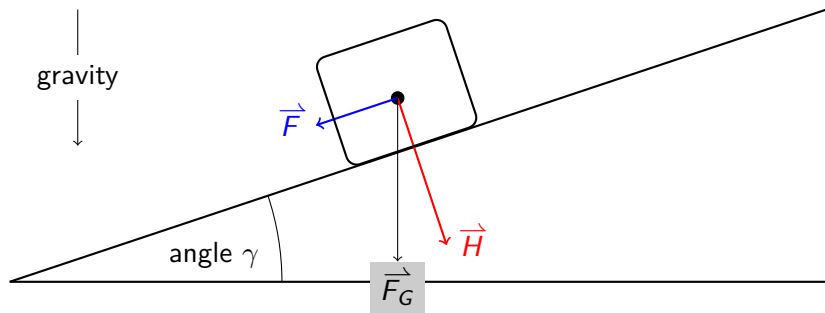
1. $\vec{x} \in W^\perp$ if and only if $\vec{x} \cdot u_i = 0$ for all $1 \leq i \leq p$
2. W^\perp is a subspace of \mathbb{R}^n
3. $(W^\perp)^\perp = W$

Examples associated to a $m \times n$ -matrix A

$$(\text{Row}A)^\perp = \text{Nul}(A) \text{ and } (\text{Col}A)^\perp = \text{Nul}A^T$$

Example: Splitting forces in physics

We know \vec{F}_G and want to split the force into two orthogonal components:



Develop **orthogonal projections** to achieve this.

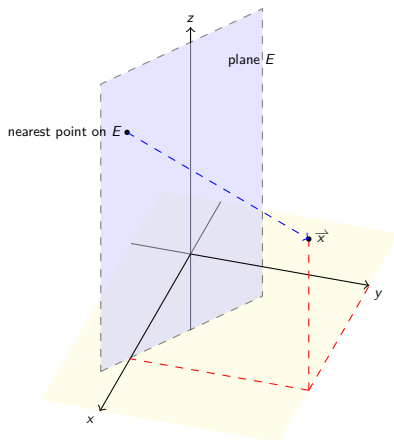
Nearest points on subspaces

The result of the orthogonal projection is the “nearest point” on a given subspace in \mathbb{R}^n .

Let $\vec{x} \in \mathbb{R}^3$,
 $E \subseteq \mathbb{R}^n$ subspace

Nearest point on E :
The point \vec{v} on E ,
with minimal distance to \vec{x}

Observe: $\vec{x} - \vec{v}$ is
perpendicular to E , i.e.
 $\vec{x} - \vec{v} \in E^\perp$



21.7 The Gram-Schmidt Process

Let $W = \text{span}\{\vec{x}_1, \dots, \vec{x}_p\} \subseteq \mathbb{R}^n$ and $\vec{x}_i \neq \vec{0}$ for $1 \leq i \leq p$.

Define

$$\vec{u}_1 = \vec{x}_1$$

$$\vec{u}_2 = \vec{x}_2 - \frac{\vec{x}_2 \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \vec{u}_1 \quad \text{if } \vec{u}_2 = \vec{0} \text{ discard and construct}$$

with same formula, now for \vec{x}_3

$$\vdots = \vdots \quad \vdots \quad \quad \quad \vdots$$

$$\vec{u}_p = \vec{x}_p - \sum_{i=1}^{p-1} \frac{\vec{x}_p \cdot \vec{v}_i}{\vec{v}_i \cdot \vec{v}_i} \vec{v}_i \quad \text{if } \vec{u}_p = \vec{0} \text{ discard}$$

Then $\{\vec{u}_1, \dots, \vec{u}_d\}$ is an orthogonal basis for W (for some $1 \leq d \leq p$) and

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