

Complementary subspaces

Let X and Y be subspaces of V , then X and Y are said to be complementary subspaces if

$$V = X + Y \text{ and } X \cap Y = 0$$

(i.e. V is a direct sum of X & Y [write $V = X \oplus Y$]).

Thm: Let B_X and B_Y be bases of X and Y , then the following are equivalent:

(1) $V = X \oplus Y$

(2) for each $v \in V$ there are unique vectors $x \in X$ and $y \in Y$ s.t. $v = x + y$.

(3) $B_X \cap B_Y = \emptyset$ and $B_X \cup B_Y$ is a basis for V .

Prove (1) \rightarrow (2)

$$\dim V = \dim(X + Y) = \dim(X) + \dim(Y) - \dim(X \cap Y)$$

$$\text{if } V = X \oplus Y \text{ \& } X \cap Y = 0 \Rightarrow$$

$$\dim V = \dim X + \dim Y$$

let $v \in V$ and assume that there are two representations for $v \Rightarrow v = x_1 + y_1 = x_2 + y_2$ $x_1, x_2 \in X; y_1, y_2 \in Y$

$$\Rightarrow x_1 - x_2 = y_2 - y_1 = \left. \begin{array}{l} x_1 - x_2 \in X \\ + \\ \cancel{x_1 - x_2} \in Y \end{array} \right\} \Rightarrow x_1 - x_2 \in X \cap Y$$

But $X \cap Y = 0 \Rightarrow x_1 = x_2$ & $y_1 = y_2$.

(2) \rightarrow (3)

$V = X + Y$ and we know that

$$B_X \cup B_Y \text{ spans } X + Y \Rightarrow$$

$$B_X \cup B_Y \text{ spans } V.$$

to prove the linearly independent, let

$$B_x = \{x_1, x_2, \dots, x_r\} \text{ \& } B_y = \{y_1, y_2, \dots, y_s\}$$

$$\Rightarrow 0 = \sum_{i=1}^r \alpha_i x_i + \sum_{j=1}^s \beta_j y_j \quad (\text{But } 0 = 0 + 0)$$

$$\Rightarrow \sum_{i=1}^r \alpha_i x_i = 0 \text{ \& } \sum_{j=1}^s \beta_j y_j = 0$$

$$\Rightarrow \alpha_1 = \dots = \alpha_r = 0 \text{ \& } \beta_1 = \beta_2 = \dots = \beta_s = 0.$$

because B_x \& B_y are linearly independent, and hence basis for V .

(3) \rightarrow (1) If $B_x \cup B_y$ is a basis for V , then

$B_x \cup B_y$ is a linearly independent set \& $B_x \cup B_y$ spans $X + Y$

$\Rightarrow B_x \cup B_y$ basis for $X + Y$ as well as V .

$$\Rightarrow V = X + Y$$

$$\dim X + \dim Y = \dim V = \dim(X + Y) \stackrel{\text{zero}}{=} \dim X + \dim Y - \dim(X \cap Y)$$

Example: let X and Y be subspaces of \mathbb{R}^3 that spanned by

$$B_x = \left\{ \begin{pmatrix} 1 \\ -1 \\ -1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ -2 \end{pmatrix} \right\} \text{ \& } B_y = \left\{ \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} \right\}$$

Explain why X and Y are complementary.

Solution: B_x and B_y are linearly independent so they are bases for X and Y ,

$$\text{rank}[X \ Y] = \text{rank} \begin{pmatrix} 1 & 0 & 1 \\ -1 & 1 & -1 \\ -1 & -2 & 0 \end{pmatrix} = 3$$

$\Rightarrow X + Y$ are complementary
 $\& B_X \cup B_Y$ bases for \mathbb{R}^3 .

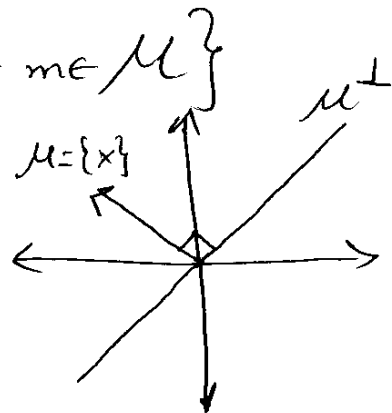
Orthogonal Complement

Def: For a subset \mathcal{M} of an inner-product space V , the orthogonal complement \mathcal{M}^\perp (read " \mathcal{M} perp") of \mathcal{M} is defined to be the set of all vectors in V that are orthogonal to every vector in \mathcal{M} , (i.e.)

$$\mathcal{M}^\perp = \{ x \in V \mid \langle m, x \rangle = 0 \quad \forall m \in \mathcal{M} \}$$

Example: (1) $\mathcal{M} = \{x\}$ in \mathbb{R}^2

\mathcal{M}^\perp will be the line through the origin that is perpendicular to x .



(2) $\mathcal{M} =$ plane through the origin

$\mathcal{M}^\perp =$ line through the origin that is perpendicular to the plane.

Note: \mathcal{M}^\perp is a subspace of V even if \mathcal{M} is not.
 (Proof exc.)

Orthogonal Complementary subspaces?

If \mathcal{M} is a subspace of a finite-dimensional inner product space V , then $V = \mathcal{M} \oplus \mathcal{M}^\perp$. Furthermore if \mathcal{N} is a subspace s.t. $V = \mathcal{M} \oplus \mathcal{N}$ and $\mathcal{N} \perp \mathcal{M}$ (every vector in \mathcal{N} is \perp to every vector in \mathcal{M}), then

$$\mathcal{N} = \mathcal{M}^\perp.$$

Proof Observe that $\mathcal{M} \cap \mathcal{M}^\perp = \{0\}$ because if $x \in \mathcal{M}$ & $x \in \mathcal{M}^\perp$, then $\langle x | x \rangle = 0 \Rightarrow x = 0$ to prove that $\mathcal{M} \oplus \mathcal{M}^\perp = V$?

Suppose that $B_{\mathcal{M}}$ & $B_{\mathcal{M}^\perp}$ are orthonormal bases for \mathcal{M} & \mathcal{M}^\perp & $\mathcal{M} \cap \mathcal{M}^\perp = \{0\}$.

$\Rightarrow B_{\mathcal{M}} \cup B_{\mathcal{M}^\perp}$ is orthonormal basis for some subspace

$$S = \mathcal{M} \oplus \mathcal{M}^\perp \subseteq V$$

If $S \neq V$ then $B_{\mathcal{M}} \cup B_{\mathcal{M}^\perp} \cup \varepsilon$ is orthonormal bases for V

$$\Rightarrow \varepsilon \perp B_{\mathcal{M}} \Rightarrow \varepsilon \perp \mathcal{M} \Rightarrow \varepsilon \subseteq \mathcal{M}^\perp$$

$\Rightarrow \varepsilon \subseteq \text{span}(B_{\mathcal{M}^\perp})$ impossible.

because $B_{\mathcal{M}} \cup B_{\mathcal{M}^\perp} \cup \varepsilon$ is linearly independent

$\Rightarrow \varepsilon = \text{empty set}$

$$\Rightarrow V = \mathcal{M} \oplus \mathcal{M}^\perp$$

• to prove $\mathcal{N} \perp \mathcal{M} \Rightarrow \mathcal{N} \subseteq \mathcal{M}^\perp$ & $\mathcal{M} \oplus \mathcal{M}^\perp = \mathcal{M} + \mathcal{N} = V$

$$\Rightarrow \dim \mathcal{N} = \dim V - \dim \mathcal{M} = \dim \mathcal{M}^\perp \quad \star$$

Example: let $U_{m \times m} = (U_1 \ U_2)$ be partitioned orthogonal matrix. Explain why $R(U_1)$ & $R(U_2)$ must be orthogonal complements of each other.

Proof: since $B_x \cap B_y = \emptyset$ & $B_x \cup B_y$ is a basis for $V \Rightarrow \mathbb{R}^m = R(U_1) \oplus R(U_2)$ and we know that

$$R(U_1) \perp R(U_2)$$

because the columns of U are orthonormal set.

$$\Rightarrow R(U_2) = R(U_1)^\perp.$$

Prop operation

If \mathcal{M} is a subspace of n -dimensional inner-product space, then the following are true:

(1) $\dim \mathcal{M}^\perp = n - \dim \mathcal{M}$

(2) $(\mathcal{M}^\perp)^\perp = \mathcal{M}$

Proof: (1) ~~is~~ is clear ~~that~~ since \mathcal{M} and \mathcal{M}^\perp are complement

(2) to prove (2) we show that $(\mathcal{M}^\perp)^\perp \subseteq \mathcal{M}$.

if $x \in (\mathcal{M}^\perp)^\perp \Rightarrow x = m + n \quad m \in \mathcal{M} \text{ \& } n \in \mathcal{M}^\perp$

$$\Rightarrow 0 = \langle n | x \rangle = \langle n | m + n \rangle = \langle n | m \rangle + \langle n | n \rangle$$

$$\Rightarrow \langle n | n \rangle = 0 \Rightarrow n = 0 \Rightarrow x \in \mathcal{M}.$$

$$\Rightarrow (\mathcal{M}^\perp)^\perp \subseteq \mathcal{M} \quad \text{from (1) } \dim \mathcal{M}^\perp = n - \dim \mathcal{M}$$

$$\text{\& } \dim (\mathcal{M}^\perp)^\perp = n - \dim \mathcal{M}^\perp \Rightarrow \dim (\mathcal{M}^\perp)^\perp = \dim \mathcal{M}$$

$$\Rightarrow (\mathcal{M}^\perp)^\perp = \mathcal{M} \quad \#.$$

Note: if $\langle x|y\rangle=0 \forall x \in V \Rightarrow \langle y|y\rangle=0 \Rightarrow y=0$

The four fundamental subspaces of $A_{m \times n}$:

~~1) $R(A)^\perp = N(A^T)$~~

if $x \in R(A)^\perp$ iff $\langle Ay|x\rangle=0 \Rightarrow y^T A^T x=0$

iff $\langle y|A^T x\rangle=0$ iff $A^T x=0$

iff $x \in N(A^T)$

$\Rightarrow R(A)^\perp = N(A^T)$

and prep both side gives $R(A^T) = N(A)^\perp$

Thm for every $A \in \mathbb{R}^{m \times n}$ (orthogonal decomposition Thm).

$R(A)^\perp = N(A^T)$ & $N(A)^\perp = R(A^T)$

$\Rightarrow \mathbb{R}^m = R(A) \oplus R(A)^\perp = R(A) \oplus N(A^T)$

$\mathbb{R}^n = N(A) \oplus N(A)^\perp = N(A) \oplus R(A^T)$. #

The last Thm tells us how to decompose \mathbb{R}^m & \mathbb{R}^n in terms of four fundamental subspaces of A , the orthogonal decomposition thm tell us how to decompose A itself into more basic components.

URV- Factorization

Suppose $\text{rank}(A) = r$, let

$$B_{R(A)} = \{u_1, \dots, u_r\} \quad \& \quad B_{N(A^T)} = \{u_{r+1}, \dots, u_m\}$$

be orthonormal bases for $R(A)$ & $N(A^T)$.

$$B_{R(A^T)} = \{v_1, \dots, v_r\} \quad \& \quad B_{N(A)} = \{v_{r+1}, \dots, v_n\}$$

be orthonormal bases for $R(A^T)$ & $N(A)$.

$\Rightarrow B_{R(A)} \cup B_{N(A^T)}$ & $B_{R(A^T)} \cup B_{N(A)}$ are orthonormal bases for \mathbb{R}^m and \mathbb{R}^n respectively.

Define $U_{m \times m} = [u_1 \ u_2 \ \dots \ u_m]$ & $V_{n \times n} = [v_1 \ v_2 \ \dots \ v_n]$

are orthogonal matrices.

Consider $R = U^T A V$, $r_{ij} = u_i^T A v_j$

However $u_i^T A = 0$ for $i = r+1, \dots, m$

$A v_j = 0$ for $j = r+1, \dots, n$

$$\Rightarrow R = U^T A V = \left(\begin{array}{cc|c} u_1^T A v_1 & \dots & u_1^T A v_r & \bigcirc \\ \vdots & & \vdots & \\ u_r^T A v_1 & \dots & u_r^T A v_r & \bigcirc \\ \hline & & & \bigcirc \\ & & & \bigcirc \end{array} \right)$$

$$\Rightarrow R = U^T A V = U^T \left(\begin{array}{c|c} C_{r \times r} & \bigcirc \\ \hline \bigcirc & \bigcirc \end{array} \right) V^T$$

where C is nonsingular matrix because it is $r \times r$, and

$$\text{rank}(C) = \text{rank} \begin{pmatrix} C & \bigcirc \\ \bigcirc & \bigcirc \end{pmatrix} = \text{rank}(U^T A V) = \text{rank}(A) = r.$$

this is URV-Factorization.