

Orthogonal Complement

Big Idea. The orthogonal complement U^\perp of a subspace U is the collection of all vectors which are orthogonal to every vector in U

Orthogonal Vectors

DEFINITION

The **inner product** of vectors $\vec{x}, \vec{y} \in \mathbb{R}^n$ is

$$\langle x, y \rangle = \sum_{k=1}^n x_k y_k = x_1 y_1 + \cdots + x_n y_n$$

Lets summarize various properties of the inner product:

1. The inner product is symmetric

$$\langle x|y \rangle = \langle y|x \rangle \forall \vec{x}, \vec{y} \in \mathbb{R}^n$$

2. The inner product of column vectors is the same as matrix multiplication

$$\langle \mathbf{x} | \mathbf{y} \rangle = \mathbf{x}^T \mathbf{y}$$

3. The inner product satisfies the usual distributive rule of multiplication

$$\langle \mathbf{x} | c\mathbf{y} + d\mathbf{z} \rangle = c \langle \mathbf{x} | \mathbf{y} \rangle + d \langle \mathbf{x} | \mathbf{z} \rangle$$

4. The square root of the inner product of a vector with itself is equal to the 2-norm

$$\sqrt{\langle \mathbf{x} | \mathbf{x} \rangle} = \|\mathbf{x}\|$$

5. We can also write the inner product in terms of the angle between vectors

$$\langle \mathbf{x} | \mathbf{y} \rangle = \|\mathbf{x}\| \|\mathbf{y}\| \cos \theta$$

6. Let A be an $m \times n$ matrix, let $\mathbf{u} \in \mathbb{R}^n$ and let $\mathbf{v} \in \mathbb{R}^m$. Then

$$\langle A\mathbf{u} | \mathbf{v} \rangle = \langle \mathbf{u} | A^T \mathbf{v} \rangle$$

7. Let us have $A = \begin{bmatrix} \mathbf{u}_1^T \\ \vdots \\ \mathbf{u}_m^T \end{bmatrix} \in \mathbb{R}^{m \times n}$, where $\mathbf{u}_1, \dots, \mathbf{u}_m \in \mathbb{R}^n$, that is \mathbf{u}_i^T is the i -th row of

A . Then for any $\mathbf{x} \in \mathbb{R}^n$, we have

$$A\mathbf{x} = \begin{bmatrix} \langle \mathbf{u}_1 | \mathbf{x} \rangle \\ \vdots \\ \langle \mathbf{u}_m | \mathbf{x} \rangle \end{bmatrix}$$

which follows directly from the definition of the matrix vector product

DEFINITION

Vectors are orthogonal if the inner product between them is zero $\langle \mathbf{x} | \mathbf{y} \rangle = 0$

More generally vectors $\mathbf{x}_1, \dots, \mathbf{x}_m \in \mathbb{R}^n$ are orthogonal if

$$\langle \mathbf{x}_i | \mathbf{x}_j \rangle = 0 \text{ for all } i \neq j$$

In other words each vector is orthogonal to every other vector in the set

Furthermore they are orthonormal if $\langle \mathbf{x}_i | \mathbf{x}_i \rangle = 1 = \|\mathbf{x}_i\|^2$ if each vector is orthogonal to the others and is a unit vector

Vectors \mathbf{x}, \mathbf{y} are orthogonal iff the acute angle between them is $\pi/2$ radians or 90°

THEOREM

Let the $\sum_n \mathbf{x}_n$ form an orthogonal basis, then

$$\|\mathbf{x}_1 + \dots + \mathbf{x}_m\|^2 = \|\mathbf{x}_1\|^2 + \dots + \|\mathbf{x}_m\|^2$$

This is called the Pythagorean theorem

Orthogonal subspaces

DEFINITION

Let $U_1, U_2 \subseteq \mathbb{R}^n$ be **subspaces**. Then U_1 and U_2 are orthogonal if $\langle \mathbf{x}_1 | \mathbf{x}_2 \rangle = 0$ for all $\mathbf{x}_1 \in U_1$ and $\mathbf{x}_2 \in U_2$. If U_1 and U_2 are orthogonal subspaces then we write $U_1 \perp U_2$

THEOREM

Let $\{\mathbf{u}_1, \dots, \mathbf{u}_l\}$ be a basis of a subspace $U_1 \subseteq \mathbb{R}^n$ and let $\{\mathbf{v}_1, \dots, \mathbf{v}_l\}$ be a basis of a subspace $U_2 \subseteq \mathbb{R}^n$. Then $U_1 \perp U_2 \iff \langle \mathbf{u}_i | \mathbf{v}_j \rangle = 0 \quad \forall i, j$

In other words, every \mathbf{u}_i in the U_1 basis is orthogonal to each \mathbf{v}_j in the basis U_2

EXAMPLE

Let $U_1 \subset \mathbb{R}^3$ and $U_2 \subset \mathbb{R}^3$ be 2-dimensional subspaces, planes. Is it possible that $U_1 \perp U_2$?

No!

Orthogonal Complement

DEFINITION

Let $U \subseteq \mathbb{R}^n$ be a subspace. The orthogonal complement of U is given by

$$U^\perp = \{\mathbf{x} \in \mathbb{R}^n : \langle \mathbf{x} | \mathbf{y} \rangle = 0 \forall \mathbf{y} \in U\}$$

Note

- If $U \subseteq \mathbb{R}^n$ is any subspace then $U = (U^\perp)^\perp$ and also $U \cap U^\perp = \{\vec{0}\}$
- $\{\mathbf{0}\}^\perp = \mathbb{R}^n$

THEOREM

Let $U \subseteq \mathbb{R}^n$ is a subspace, then $U^\perp \subseteq \mathbb{R}^n$ is a subspace

PROOF

Let us verify that U^\perp satisfies the properties of a subspace

clearly $\langle \vec{0} | \mathbf{x} \rangle = 0 \forall \mathbf{x} \in U$ therefore $\vec{0} \in U^\perp$

Let $\mathbf{x}_1, \mathbf{x}_2 \in U^\perp$. Then

$$\langle \mathbf{x}_1 + \mathbf{x}_2 | \mathbf{y} \rangle = \langle \mathbf{x}_1 | \mathbf{y} \rangle + \langle \mathbf{x}_2 | \mathbf{y} \rangle = 0 + 0 = 0 \forall \mathbf{y} \in U$$

Therefore $\mathbf{x}_1 + \mathbf{x}_2 \in U^\perp$

Let $c \in \mathbb{R}$, $\mathbf{u} \in U^\perp$. Then

$$\langle c\mathbf{x} | \mathbf{y} \rangle = c \langle \mathbf{x} | \mathbf{y} \rangle = c(0) = 0 \forall \mathbf{y} \in U$$

Therefore $c\mathbf{x} \in U^\perp$

Therefore U^\perp is a subspace

Fundamental subspaces

DEFINITION

Let A be a $m \times n$ matrix, The fundamental subspaces of A are $N(A)$, $R(A)$, $N(A^T)$ and $R(A^T)$

THEOREM

Let A be an $m \times n$ matrix. Then $N(A) = R(A^T)^\perp$ and $R(A) = N(A^T)^\perp$

PROOF

The *second* equality follows from the first by replacing A with A^T therefore it is sufficient to prove $N(A) = R(A^T)^\perp$

A general strategy to prove equality of sets is to show that each set contains the other therefore lets prove $N(A) \subseteq R(A^T)^\perp$ and then the reverse

Let $\mathbf{x} \in N(A)$. Then $A\mathbf{x} = \mathbf{0}$ and so $\langle A\mathbf{x} | \mathbf{y} \rangle = 0$ for all $\mathbf{y} \in \mathbb{R}^m$

Using properties of the inner product we see that $\langle \mathbf{x} | A^T \mathbf{y} \rangle = 0 \forall \mathbf{y} \in \mathbb{R}^m$
 $\therefore \mathbf{x} \in R(A^T)^\perp$

Let $\mathbf{x} \in R(A^T)^\perp$. Then $\langle \mathbf{x} | A^T \mathbf{y} \rangle = 0$ and so $\langle A\mathbf{x} | \mathbf{y} \rangle = 0 \forall \mathbf{y} \in \mathbb{R}^m$. Choose $\mathbf{y} = A\mathbf{x} \in \mathbb{R}^m$

and then $\langle A\mathbf{x} | A\mathbf{x} \rangle = 0 \implies \|A\mathbf{x}\| = 0 \implies A\mathbf{x} = \vec{0}$ and finally $\mathbf{x} \in N(A)$

THEOREM

Let $U \subseteq \mathbb{R}^n$ be a subspace. Then

$$\dim(U) + \dim(U^\perp) = n$$

PROOF

Let $\dim(U) = m$ and let $\sum_m \mathbf{u}_m$ be a basis of U and define

$$A = \begin{bmatrix} \mathbf{u}_1^T \\ \vdots \\ \mathbf{u}_m^T \end{bmatrix}$$

Then $U = R(A^T)$ and $U^\perp = R(A^T)^\perp = N(A)$ and we know $\text{rank}(A) = m = \dim(U)$
 therefore

$$\dim(U) + \dim(U^\perp) = \text{rank}(A) + \dim(N(A)) = n$$

by the Rank Nullity Theorem

EXAMPLE

Let A be a matrix such that its LU Decomposition is of the form

$$A = LU = \begin{bmatrix} 1 & 0 & 0 \\ * & 1 & 0 \\ * & * & 1 \end{bmatrix} \begin{bmatrix} * & * & * & * \\ 0 & * & * & * \\ 0 & 0 & * & * \end{bmatrix}$$

where $*$ denotes a non zero number. Find the dimension of each subspace $N(A)$, $R(A)$, $N(A^T)$ and $R(A^T)$

Clearly $\dim(N(A)) = 1$ and $\dim(R(A)) = 3$ therefore

$$\dim(N(A^T)) = \dim(R(A)^\perp) = 3 - 3 = 0$$

and

$$\dim(R(A^T)) = \dim(N(A)^\perp) = 4 - 1 = 3$$