

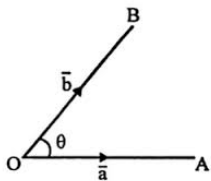
# Inner Product and Orthogonality

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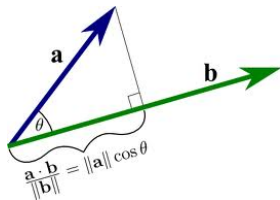
In the Euclidean space  $\mathbb{R}^2$  and  $\mathbb{R}^3$  there are two concepts, viz., length (or distance) and angle which have no analogues over a general field.

**Fortunately there is a single concept** usually known as scalar product or dot product which covers both the concepts of length and angle.



Let  $a = (a_1, a_2)$  and  $b = (b_1, b_2)$  be vectors in  $\mathbb{R}^2$  represented by the points  $A$  and  $B$  as in figure. Then the scalar product of  $a$  and  $b$  is defined to be

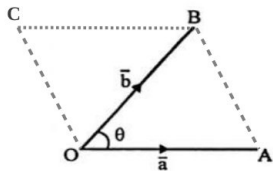
$$\langle a, b \rangle = l_1 l_2 \cos \theta$$



where  $l_1$  is the length of  $OA$ ,  $l_2$  is the length of  $OB$  and  $\theta$  is the angle between  $OA$  and  $OB$ .

It can be shown using trigonometry that  $l_1 l_2 \cos \theta = a_1 b_1 + a_2 b_2$ , so  $\langle a, b \rangle = a_1 b_1 + a_2 b_2$ . **In Linear Algebra, scalar product is called inner product.**

# Length, distance, angle : in terms of the inner product



- 1 **Length of a vector:** The length  $OA$  can be defined in terms of the inner product since

$$OA^2 = \langle a, a \rangle.$$

- 2 **Distance between vectors:** If  $OABC$  is a parallelogram, the distance  $AB = OC = \sqrt{\langle b - a, b - a \rangle}$  since  $C = b - a$ .

- 3 The angle  $\theta$  can be obtained as

$$\theta = \cos^{-1} \left( \frac{\langle a, b \rangle}{\sqrt{\langle a, a \rangle \cdot \langle b, b \rangle}} \right).$$

The above concepts and results have obvious analogues in  $\mathbb{R}^3$ .

We shall extend them to arbitrary (finite-dimensional) **vector spaces over  $\mathbb{R}$  or  $\mathbb{C}$** .

One does not extend inner product to vector spaces over a general field mainly because  $\langle x, x \rangle \geq 0$  has no meaning in a general field.

# Problems

- 1 Let  $z$  be a fixed nonnull vector in the plane. What is the locus of the point  $x$  such that  $\langle x, z \rangle = 0$ ? What happens if 0 is replaced by a non-zero scalar?
- 2 If  $x_1, x_2, y_1, y_2$  are real numbers, show that

$$(x_1x_2 + y_1y_2)^2 \leq (x_1^2 + y_1^2)(x_2^2 + y_2^2).$$

Hence deduce that  $PQ + QR \geq PR$  for any three points  $P, Q$  and  $R$  in the plane.

Motivated by the usual inner product (dot product) on  $\mathbb{R}^2$  we now give the axiomatic definition of inner product on a vector space over  $F$ , where  $F$  is either  $\mathbb{R}$  or  $\mathbb{C}$ .

### Definition

An inner product on a vector space  $V$  over  $F$  is a map  $(x, y) \mapsto \langle x, y \rangle$  from  $V \times V$  to  $F$  satisfying the following three conditions:

- 1  $\langle x, y \rangle = \overline{\langle y, x \rangle}$ .
- 2  $\langle \alpha x + \beta y, z \rangle = \alpha \langle x, z \rangle + \beta \langle y, z \rangle$
- 3  $\langle x, x \rangle \geq 0$  ;  $\langle x, x \rangle = 0 \iff x = 0$ .

a vector space with an inner product	an <b>inner product space</b>
a real inner product space	a <b>Euclidean space</b>
a complex inner product space	a <b>unitary space</b>

# Properties of an inner product

- 1 The restriction of an inner product to a subspace is an inner product.
- 2 In any inner product space, we have
  - $\langle x, \alpha y + \beta z \rangle = \bar{\alpha} \langle x, y \rangle + \bar{\beta} \langle x, z \rangle$ .
  - $\langle 0, y \rangle = \langle x, 0 \rangle = 0$ .
- 3 When the second argument is held fixed, **inner product is linear in the first argument**. Similarly, when the first argument is held fixed, **inner product is conjugate-linear in the second argument**.



# Examples of an inner product

- 1 The inner product

$$\langle x, y \rangle = \sum_{i=1}^n x_i y_i = y^T x$$

on  $\mathbb{R}^2$  is called the **canonical inner product** on  $\mathbb{R}^n$ .

- 2 On  $\mathbb{C}^n$ , the **canonical inner product** is defined by

$$\langle x, y \rangle = \sum_{i=1}^n x_i \bar{y}_i = y^* x.$$

$y^*$ , the **adjoint** of  $y$ , to denote  $\bar{y}^T$ .

- 3 Fix any finite subset  $A$  of  $\mathbb{R}$  with size  $\geq n$ . Let  $V = \mathcal{P}_n$  over  $\mathbb{R}$ .

$$\langle p, q \rangle := \sum_{a \in A} p(a)q(a)$$

is an inner product on  $V$ .

- 4  $\langle A, B \rangle = \text{tr}(B^* A)$  is an inner product on  $\mathbb{C}^{m \times n}$ .

## Examples of an inner product

- 1 On the vector space  $V$  of all real-valued continuous functions on an interval  $[a, b]$ ,

$$\langle f, g \rangle = \int_a^b f(t) g(t) dt$$

defines an inner product.

- 2 If  $h \in V$  is such that  $h(t) > 0$  for all  $t \in [a, b]$ ,

$$\langle f, g \rangle = \int_a^b h(t) f(t) g(t) dt$$

is also an inner product.

- 3 Let  $V$  be the vector space of all real-valued random variables with mean 0 and finite variance, defined on a fixed probability space. Let  $F = \mathbb{R}$  and define  $\langle x, y \rangle$  to be the covariance between  $x$  and  $y$ .

## Examples of non-inner product spaces

- 1  $\langle x, y \rangle := y^T x$  and  $\langle x, y \rangle := x^* y$  are **not** inner products on  $\mathbb{C}^n$ .
- 2  $\langle A, B \rangle = \sum_{i=1}^n a_{ii} \bar{b}_{ii}$  is **not** an inner product on  $\mathbb{C}^{m \times n}$ .  
What are all the axioms which are violated?

## Inner product associated with a matrix

Let  $V$  be an inner product space and  $\mathcal{B} = \{u_1, u_2, \dots, u_n\}$  a basis of  $V$ . Let  $\alpha = (\alpha_1, \alpha_2, \dots, \alpha_n)^T$  and  $\beta = (\beta_1, \beta_2, \dots, \beta_n)^T$  be the coordinate vectors of  $x$  and  $y$  respectively with respect to  $\mathcal{B}$  and let  $A = (a_{ij})$ , where  $a_{ij} = \langle u_j, u_i \rangle$ . Then

$$\langle x, y \rangle := \left\langle \sum \alpha_i u_i, \sum \beta_j u_j \right\rangle = \sum \sum \overline{\beta_j} a_{ji} \alpha_i = \beta^* A \alpha. \quad (1)$$

The matrix  $A$  will satisfy the following conditions:

- 1  $A = A^*$
- 2  $\alpha^* A \alpha \geq 0$  for all  $\alpha \in F^n$ ,
- 3 if  $\alpha^* A \alpha = 0$  then  $\alpha = 0$ .

## Matrix associated with an inner product

Conversely, if  $A$  is a matrix satisfying the above three conditions, then  $\langle \cdot, \cdot \rangle$  defined by (1) is an inner product on  $V$ .

Suppose  $A = B^*B$ , where  $B$  is a matrix with  $n$  columns and rank  $n$ . Then

$$\langle x, y \rangle = y^* B^* B x$$

is an inner product because

- $\overline{\langle y, x \rangle} = \overline{x^* B^* B y} = (x^* B^* B y)^* = y^* B^* B x = \langle x, y \rangle$ .
- $\langle x, x \rangle = (Bx)^*(Bx) \geq 0$ .
- If  $\langle x, x \rangle = 0$  then  $Bx = 0$  and so  $x = 0$ .

We shall later show that any matrix  $A$  satisfying the above three conditions, can be written as  $B^*B$  for some non-singular  $B$ .

## Concept of length : Norm

**Inner** combines the concepts of length and angle. We shall discuss the first concept, length.

### Definition

A **norm** on a (real or complex) vector space  $V$  is a map  $x \mapsto \|x\|$  from  $V$  to  $\mathbb{R}$  satisfying the following three conditions:

- 1  $\|x\| \geq 0$  ;  $x = 0$  if  $\|x\| = 0$
- 2  $\|\alpha x\| = |\alpha| \cdot \|x\|$
- 3  $\|x + y\| \leq \|x\| + \|y\|$ .

A vector space together with a norm on it is called a **normed vector space** or **normed linear space**.

We shall prove that every inner product induces a norm. We shall give a family of norms which are not induced by inner product. For this we need some famous inequalities.

## Concept of angle

Inner combines the concepts of length and angle. We shall discuss an important special case of the second concept, viz., the angle between two vectors being  $90^\circ$ .

Let  $V$  be an inner product space,  $x, y \in V$ . Let  $A, B$  be subsets of  $V$ .

$\langle x, y \rangle = 0$ (we write $x \perp y$ )	$x$ and $y$ are <b>orthogonal</b> to each other
$x \perp y$ for every pair of distinct vectors $x, y$ in $A$	$A$ is <b>orthogonal</b>
$A$ is orthogonal and every vector in $A$ has norm 1	$A$ is <b>orthonormal</b>
every vector in $A$ is orthogonal to every vector in $B$	$A$ is <b>orthogonal</b> to $B$



- 1  $x \perp y \iff y \perp x$ .
- 2  $0 \perp x$  for all  $x$ .
- 3  $x \perp x \iff x = 0$ .
- 4 if  $x \perp y, y \perp z$ , then  $x \perp (\alpha y + \beta z)$ .
- 5 A set of vectors is orthogonal iff its elements are pair-wise orthogonal.  
**Is the corresponding statement for linear independence true?**  
Linear independence is a property of the entire set whereas orthogonality is a property of pairs.
- 6 The empty set is orthonormal (in a vacuous sense).

# Pythagoras theorem

In a **real inner product space**, if  $x \perp y$ , then

$$\|x + y\|^2 = \|x\|^2 + \|y\|^2.$$

The converse is true for real inner product space but not for complex inner product space.

More generally,

$$\left\| \sum_{i=1}^k x_i \right\|^2 = \sum_{i=1}^k \|x_i\|^2$$

if  $\{x_1, x_2, \dots, x_k\}$  is orthogonal. The converse is not true for both real and complex inner product spaces.

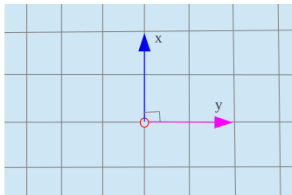
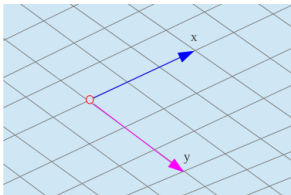
- 1 Any orthogonal set  $A$  not containing the null vector is linearly independent.
- 2 Any orthonormal set is linearly independent.
- 3 If the subspaces  $S_1, S_2, \dots, S_k$  are orthogonal to one another then  $S_1 + S_2 + \dots + S_k$  is direct.

## Definition

Let  $S$  be a subspace of an inner product space. We say that  $B$  is an **orthogonal basis** (resp. an **orthonormal basis**) of  $S$  if  $B$  is a basis of  $S$  and  $B$  is an orthonormal (resp. an orthonormal) set.

We have seen that a basis corresponds to a **coordinate system**.

An orthonormal basis corresponds to a **system of rectangular coordinates** where the reference point on each axis is at unit distance from the origin.



For a given orthonormal basis, finding the coordinates with respect to such a coordinate system is easy as shown in the following.

### Theorem

Let  $B = \{x_1, x_2, \dots, x_n\}$  be an orthonormal basis of an inner product space  $V$ . Then for any  $x \in V$ , we have

$$s = \sum_{j=1}^n \langle x, x_j \rangle x_j.$$

Suppose  $A = \{x_1, x_2, \dots, x_k\}$  be an orthogonal set (not a basis) of non-null vectors in  $V$ . Then for any  $x \in V$ , we call

$$z := x - \sum_{j=1}^k \frac{\langle x, x_j \rangle}{\langle x_j, x_j \rangle} x_j$$

the **residual of  $x$  with respect to  $A$** . The residual  $z$  is orthogonal to each  $x_j$ .

## Exercise.

Let  $x_1, x_2, \dots, x_k$  form an orthonormal set.

- 1 Show that  $\|\sum_{i=1}^k \alpha_i x_i\|^2 = \sum_{i=1}^k \|\alpha_i\|^2$ .
- 2 If  $z$  is the residual of  $x$  on  $\{x_1, x_2, \dots, x_k\}$ , show that

$$\|z\|^2 = \|x\|^2 - \left\| \sum_{i=1}^k \langle x, x_i \rangle x_i \right\|^2 = \|x\|^2 - \sum_{i=1}^k |\langle x, x_i \rangle|^2.$$

- 3 **Bessel's inequality:**

$$\|x\|^2 \geq \sum_{i=1}^k |\langle x, x_i \rangle|^2$$

for any  $x$ . Show also that equality holds iff  $x \in \text{Sp}(\{x_1, x_2, \dots, x_k\})$ .

## Exercise

Let  $B = \{x_1, x_2, \dots, x_k\}$  be an orthonormal set in a finite-dimensional inner product space  $V$ . Show that the following statements are equivalent:

- 1  $B$  is maximal,
- 2  $\langle x, x_i \rangle = 0$  for  $i = 1, 2, \dots, k \Rightarrow x = 0$ ,
- 3  $B$  generates  $V$ ,
- 4 if  $x \in V$  then  $x = \sum_{i=1}^k \langle x, x_i \rangle x_i$ ,
- 5 if  $x, y \in V$  then  $\langle x, y \rangle = \sum_{i=1}^k \langle x, x_i \rangle \cdot \langle x_i, y \rangle$ ,
- 6 if  $x \in V$  then  $\|x\|^2 = \sum_{i=1}^k |\langle x, x_i \rangle|^2$ .

## Gram-Schmidt orthogonalization process

### Theorem

Let  $\{x_1, x_2, \dots, x_k\}$  be a basis of a subspace  $S$  of an inner product space  $V$ . Define  $z_1, z_2, \dots, z_k$  inductively by :

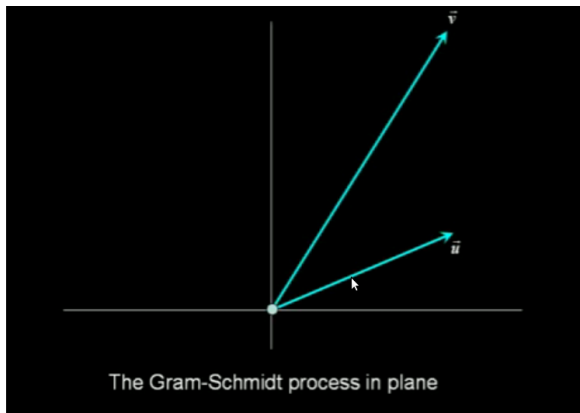
$$z_i = x_i - \sum_{j=1}^{i-1} \frac{\langle x_i, z_j \rangle}{\langle z_j, z_j \rangle} z_j \quad (i = 1, 2, \dots, k).$$

Then  $z_1, z_2, \dots, z_k$  is an **orthogonal basis** of  $S$ .

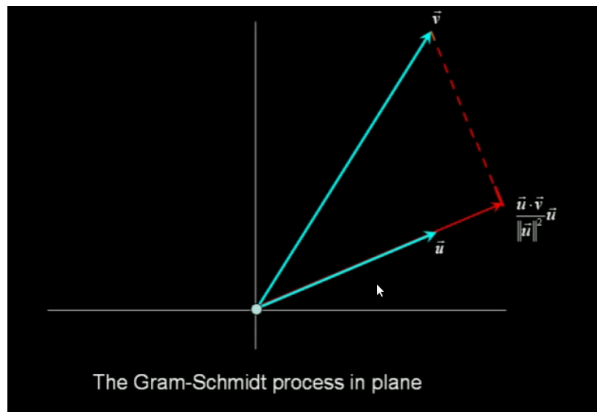
- An orthonormal basis of  $S$  can be obtained by normalizing the  $z_i$ 's.
- Note that each  $z_i$  is the residual of  $x_i$  with respect to  $z_1, z_2, \dots, z_{i-1}$  ;  $z_1, z_2, \dots, z_k$  and  $x_1, x_2, \dots, x_k$  have the same span.
- Let  $S$  be a subspace of a finite-dimensional inner product space  $V$ . Starting from any basis of  $S$  we can construct an orthonormal basis by the Gram-Schmidt process: **Every subspace  $S$  of a finite-dimensional inner product space has an orthonormal basis.**



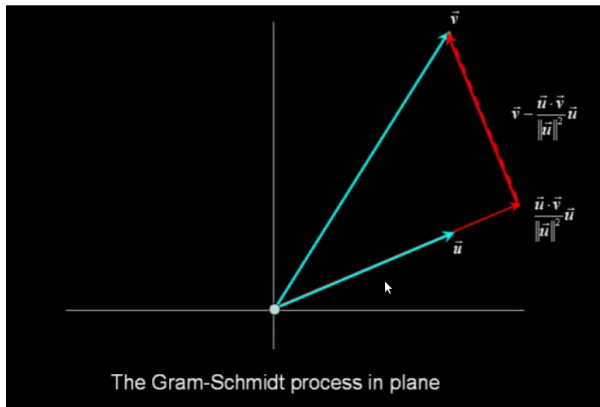
# Gram-Schmidt process in plane



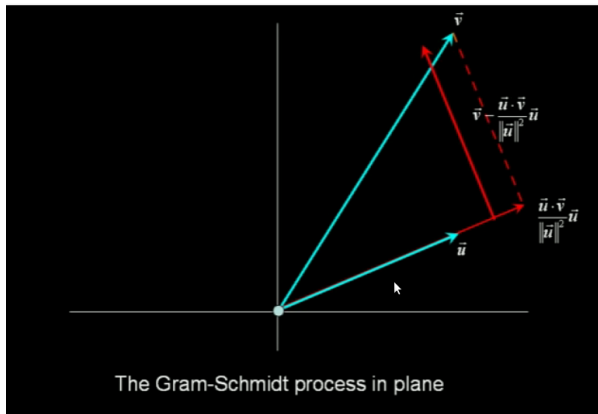
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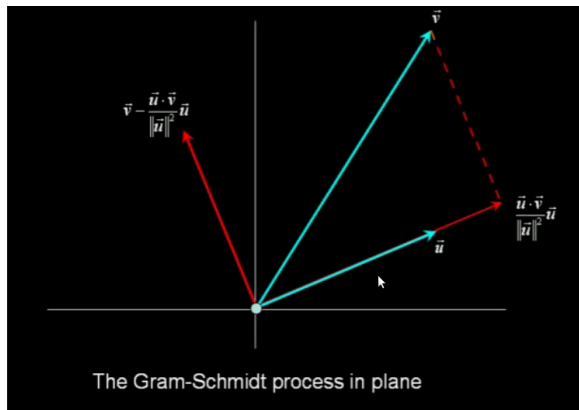
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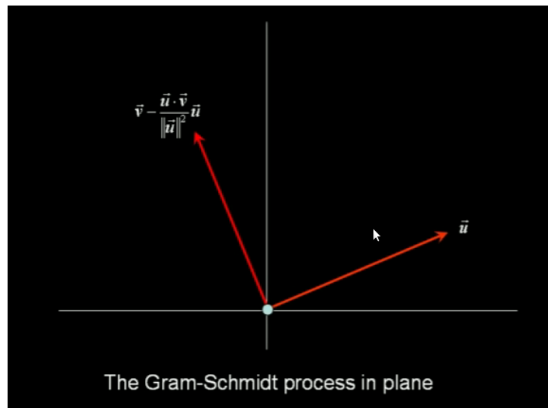
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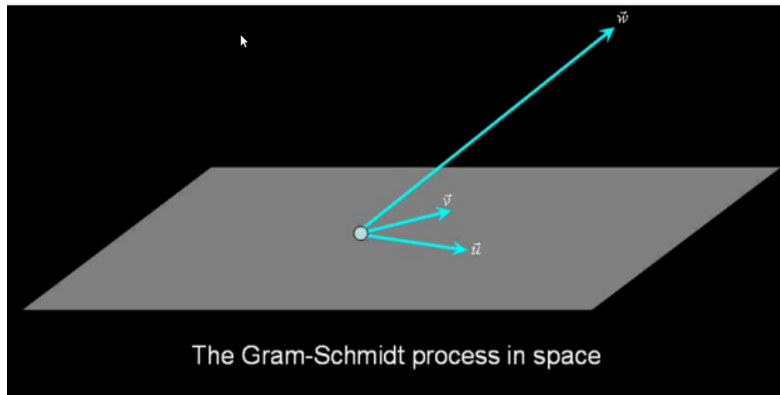
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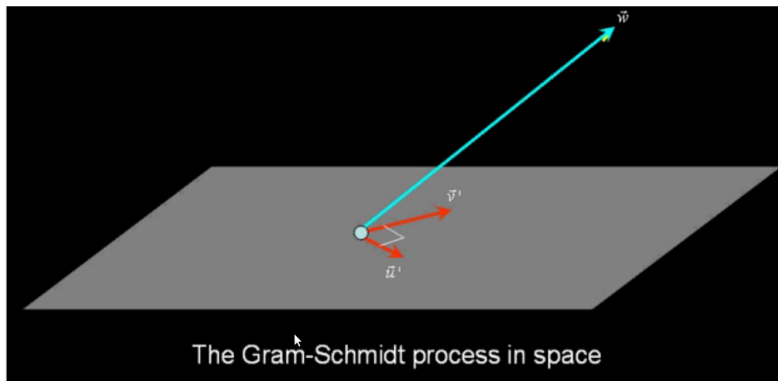
# Gram-Schmidt process in plane



# Gram-Schmidt process in space

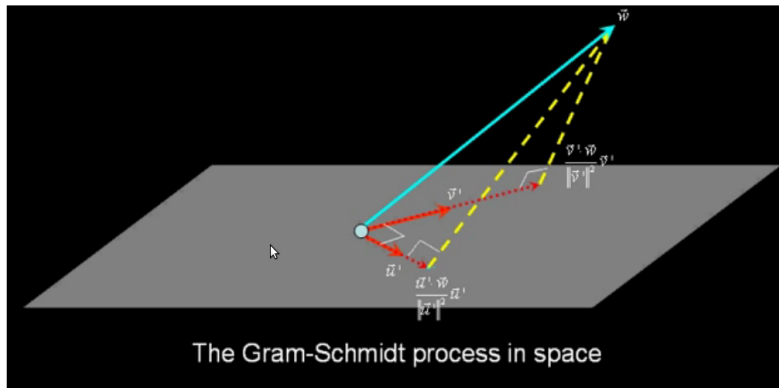


# Gram-Schmidt process in space

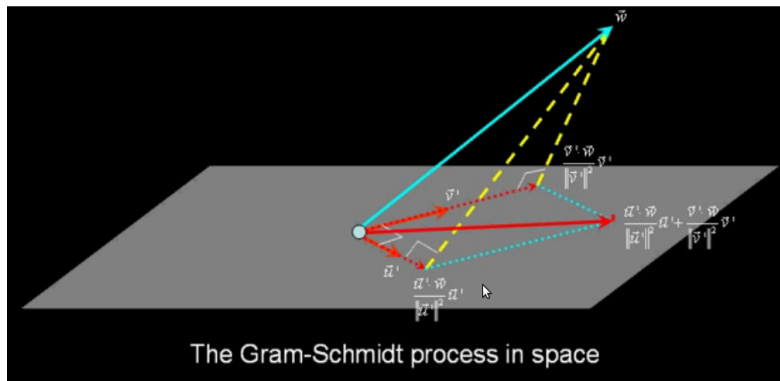




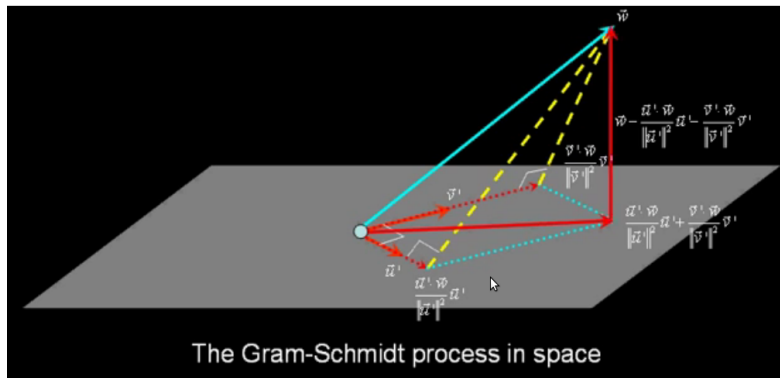
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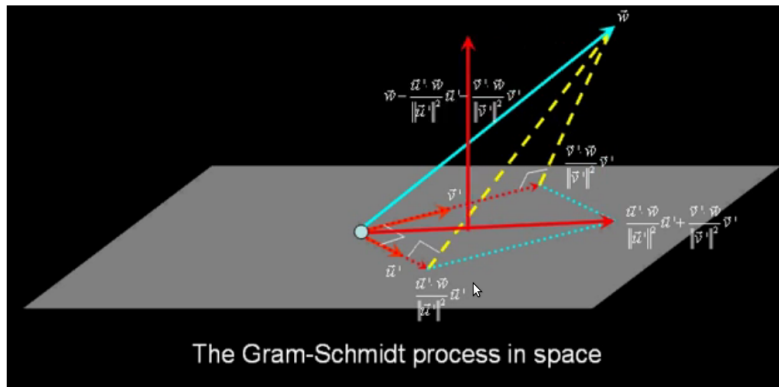
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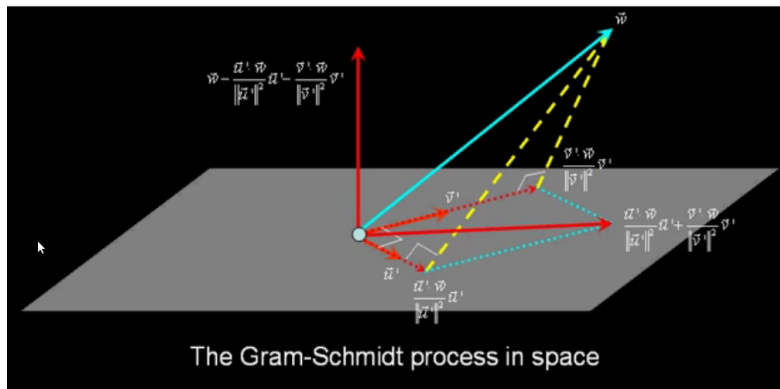
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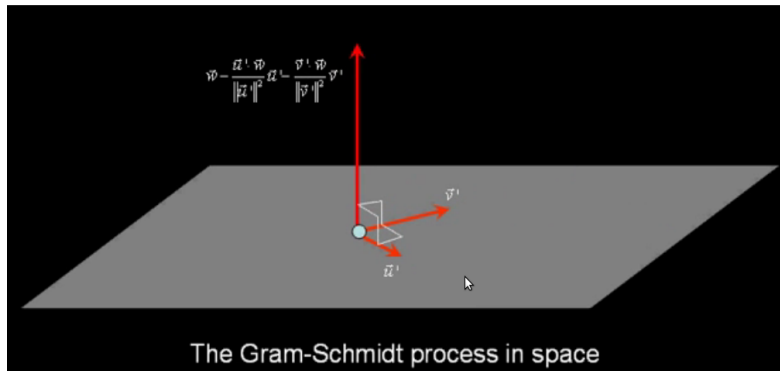
# Gram-Schmidt process in space



# Gram-Schmidt process in space



# Gram-Schmidt process in space



# Generalized Gram-Schmidt Process

Let  $x_1, x_2, \dots, x_s$  be a given vectors in  $V$ , not necessarily basis.

- 1 **Step 1:** Set  $k = 1$ .
- 2 **Step 2:** Compute  $z_k = x_k - \sum_{j=1}^{k-1} \frac{\langle x_k, y_j \rangle}{\langle y_j, y_j \rangle} y_j$ .
- 3 **Step 3:** Compute  $y_k := \frac{z_k}{\|z_k\|}$  or 0 according as  $z_k \neq 0$  or  $z_k = 0$ .
- 4 **Step 4:** If  $k < s$ , increase  $k$  by 1 and go to Step 2. Otherwise go to Step 5.
- 5 **Step 5:** For  $i = 1, 2, \dots, s$ , the set  $B_i$  of all non-null vectors **among**  $y_1, y_2, \dots, y_i$  is an orthonormal basis of the span  $S_i$  of  $\{x_1, x_2, \dots, x_i\}$ .

If  $x_1, x_2, \dots, x_\ell$  form an orthonormal set then  $y_j = x_j$  for  $j = 1, 2, \dots, \ell$ .

## Theorem

Let  $S$  be a subspace of a finite-dimensional inner product space  $V$ . Any orthonormal subset of  $S$  can be extended to an orthonormal basis of  $S$ .

**Proof.** Let  $A = \{x_1, x_2, \dots, x_\ell\}$  be an orthonormal subset of  $S$ . Extend  $A$  to a spanning set  $\{x_1, x_2, \dots, x_\ell, x_{\ell+1}, \dots, x_s\}$  of  $S$  by **appending a basis**. Applying the generalised Gram-Schmidt process to  $\{x_1, x_2, \dots, x_s\}$ , get  $\{y_1, y_2, \dots, y_s\}$ . Then the non-null vectors among  $y_1, y_2, \dots, y_s$  form an orthonormal basis of  $S$  which contains  $A = \{x_1, x_2, \dots, x_\ell\}$ .

**We note that the orthonormal basis obtained by the Gram-Schmidt process from  $x_1, x_2, \dots, x_\ell$  may be quite different from that obtained from generalised Gram-Schmidt process (a rearrangement of  $x_1, x_2, \dots, x_\ell$ ).**



## QR-decomposition.

Let  $A$  be an  $n \times s$  matrix with rank  $p$ . Let  $y_1, y_2, \dots, y_s$  be the vectors obtained when generalized Gram-Schmidt process is applied to the columns of  $A$ . For each  $k = 1, 2, \dots, s$ ,

$$z_k = A_{*k} - \sum_{j=1}^{k-1} \langle A_{*k}, y_j \rangle y_j = \|z_k\| y_k$$

and,  $y_k := \frac{z_k}{\|z_k\|}$  or 0 according as  $z_k \neq 0$  or  $z_k = 0$ .

Hence  $k$ -th column of  $A$  is a linear combination of  $y_1, y_2, \dots, y_s$ . That is,

$$A_{*k} = \langle A_{*k}, y_1 \rangle y_1 + \langle A_{*k}, y_2 \rangle y_2 + \cdots + \langle A_{*k}, y_{k-1} \rangle y_{k-1} + \|z_k\| y_k.$$

## QR-decomposition.

$$A = [y_1 \ y_2 \ \cdots \ y_s] \begin{bmatrix} \|z_1\| & \langle A_{*2}, y_1 \rangle & \langle A_{*3}, y_1 \rangle & \cdots & \langle A_{*s}, y_1 \rangle \\ 0 & \|z_2\| & \langle A_{*3}, y_2 \rangle & \cdots & \langle A_{*s}, y_2 \rangle \\ 0 & 0 & \|z_3\| & \cdots & \langle A_{*s}, y_3 \rangle \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & \|z_s\| \end{bmatrix}.$$

Let  $U$  be the  $s \times s$  upper triangular matrix ( $u_{ik}$ ) where

$$u_{ik} = \begin{cases} \langle A_{*k}, y_i \rangle & \text{if } i < k \\ \|z_k\| & \text{if } i = k \\ 0 & \text{otherwise} \end{cases}$$

Then  $A = PU$ .

Also if  $Q$  is the submatrix of  $P$  formed by the non-null columns (the columns of  $Q$  form an orthonormal basis,  $Q^*Q = I_p$ ) and  $R$  the submatrix of  $U$  formed by the corresponding rows, then  $(Q, R)$  is a rank-factorization of  $A$  and  $Q^*Q = I_p$ .

When  $A$  is of full column rank  $(Q, R) = (P, U)$  is known as a **QR-decomposition of  $A$** .

**Uniqueness.**  $QR$ -factorization is unique if **we insist that the diagonal elements of  $R$  are real and positive**, i.e., if  $A$  is of full column rank, then there exist unique matrices  $Q$  and  $R$  such that  $A = QR$ ,  $Q^*Q = I$ ,  $R$  is upper triangular and  $r_{ii} > 0$  for all  $i$ .

# Exercises

- 1 Let  $x, y, u$  and  $v$  belong to  $\mathbb{R}^n$ . Then show that

$$\langle x + iy, u + iv \rangle := u^T x + v^T y$$

is an inner product on the vector space  $\mathbb{C}^n$  over  $\mathbb{R}$ .

What is its connection with the canonical inner product on  $\mathbb{C}^n$ ?

- 2 Show that  $\langle x, y \rangle = 0$  for all  $y$  iff  $x = 0$ .

# References

- S. Kumaresan, “*Linear Algebra - A Geometric Approach*”, PHI Learning Pvt. Ltd., 2011.
- A. Ramachandra Rao and P. Bhimasankaram, “*Linear Algebra*”, Hindustan Book Agency, 2000.