

Last time:

- $u \cdot v = u_1 v_1 + u_2 v_2 + \dots + u_n v_n$
- u is orthogonal to v if $u \cdot v = 0$
- u is orthogonal to the subspace W if $u \cdot w = 0$ for all $w \in W$.
- the orthogonal complement of a subspace W
is the subspace $W^\perp = \{x \in V \mid x \cdot w = 0 \text{ for all } w \in W\}$

Theorem $(\text{Row } A)^\perp = \text{Null } A$.

Corollary If $V \subset \mathbb{R}^n$, $\dim V + \dim V^\perp = n$.

Today: a set of vectors is orthogonal if
the elements are pairwise orthogonal.

Example.

Consider

$$U_1 = \begin{bmatrix} 3 \\ -2 \\ 1 \\ 3 \end{bmatrix}, U_2 = \begin{bmatrix} -1 \\ 3 \\ -3 \\ 4 \end{bmatrix}, U_3 = \begin{bmatrix} 3 \\ 8 \\ 7 \\ 0 \end{bmatrix}$$

Show that $\{U_1, U_2, U_3\}$
is an orthogonal set.

We need to verify that

$$U_1 \cdot U_2 = 0, \quad U_1 \cdot U_3 = 0, \quad \text{and} \quad U_2 \cdot U_3 = 0.$$

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$$U_1 \cdot U_2 = -3 + -6 + -3 + 12 = 0$$

etc.

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Theorem

If $S = \{v_1, \dots, v_k\}$ is an orthogonal set of nonzero vectors, then S is a linearly independent set.
 (and hence a basis for $\text{span } S$)

Proof

Suppose $c_1v_1 + \dots + c_kv_k = 0$ for some scalars c_i .

$$\begin{aligned} \text{Then } 0 &= (c_1v_1 + \dots + c_kv_k) \cdot v_i \\ &= c_1(v_i \cdot v_i) + \cancel{c_2(v_2 \cdot v_i)}^{\rightarrow 0} + \dots + \cancel{c_k(v_k \cdot v_i)}^{\rightarrow 0} \\ &= c_1(v_i \cdot v_i). \end{aligned}$$

Since v_i is nonzero, $v_i \cdot v_i$ is nonzero, so we conclude $c_1 = 0$.

A similar argument shows that each of the c_i are zero.

□

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Definition

An orthogonal basis for a subspace W of \mathbb{R}^n
is a basis of W that is an orthogonal set.

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Example

Given

$$U_1 = \begin{bmatrix} 1 \\ 2 \\ 1 \\ 0 \end{bmatrix}, U_2 = \begin{bmatrix} 1 \\ -1 \\ 1 \\ 3 \end{bmatrix}, U_3 = \begin{bmatrix} 2 \\ -1 \\ 0 \\ -1 \end{bmatrix},$$

find a nonzero vector

$$U_4 = \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \text{ so}$$

$\{U_1, U_2, U_3, U_4\}$ forms
an orthogonal set.

(and hence a basis for \mathbb{R}^4 !)

We are looking for a vector satisfying

$$\begin{bmatrix} 1 \\ 2 \\ 1 \\ 0 \end{bmatrix} \cdot \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \\ 1 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = \begin{bmatrix} 2 \\ -1 \\ 0 \\ -1 \end{bmatrix} \cdot \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = 0,$$

i.e. a soln to the linear homogeneous equations

$$a + 2b + c = 0$$

$$a - b + c + 3d = 0$$

$$2a - b - d = 0,$$

i.e. something in the null space of

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

(Notice we are looking for something in
 $\{U_1, U_2, U_3\}^\perp = (\text{Row } A)^\perp = \text{Null } A$.)

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row-reducing, $\rightsquigarrow \begin{bmatrix} 1 & 0 & 0 & -1 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & 3 \end{bmatrix}$

so the general solution has d free, and $a=b=d$, $c=-3d$.

These solutions are spanned by $\begin{bmatrix} 1 \\ -3 \\ 1 \end{bmatrix}$ and we can

use that as our \mathbf{c}_4 .

Theorem

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Let $\{v_1, \dots, v_k\}$ be an orthogonal basis for a subspace W of \mathbb{R}^n .

Then for any vector $w \in W$, the unique scalars c_1, \dots, c_k so that $w = c_1 v_1 + \dots + c_k v_k$

are given by the formula

$$c_i = \frac{w \cdot v_i}{v_i \cdot v_i}$$

(That sure beats row-reducing!)

Proof Since we have a basis,

$$w = c_1 v_1 + \dots + c_k v_k \text{ for some scalars } c_i.$$

To compute c_i , we take the inner product with v_i :

$$\begin{aligned} w \cdot v_i &= c_1(v_i \cdot v_i) + c_2(v_2 \cdot v_i) + \dots + c_k(v_k \cdot v_i) \\ &= c_i(v_i \cdot v_i) \end{aligned}$$

and rearranging:

$$c_i = \frac{w \cdot v_i}{v_i \cdot v_i}.$$

This works for all the c_i , giving our formula.

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Example

Given the orthogonal basis

$$U = \left\{ \begin{bmatrix} 3 \\ -3 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 2 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 \\ 4 \\ 4 \end{bmatrix} \right\}$$

for \mathbb{R}^3 , find the
U-coordinates of

$$x = \begin{bmatrix} 4 \\ 2 \\ -1 \end{bmatrix}.$$

Exercise: Verify U is actually orthogonal.

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We compute

$$x \cdot u_1 = 12 - 6 + 0 = 6$$

$$x \cdot u_2 = 8 + 4 + 1 = 13$$

$$x \cdot u_3 = 4 + 2 - 4 = 2$$

$$u_1 \cdot u_1 = 18, \quad u_2 \cdot u_2 = 9, \quad u_3 \cdot u_3 = 18.$$

$$\begin{aligned} \text{Then } x &= \frac{x \cdot u_1}{u_1 \cdot u_1} u_1 + \frac{x \cdot u_2}{u_2 \cdot u_2} u_2 + \frac{x \cdot u_3}{u_3 \cdot u_3} u_3 \\ &= \frac{6}{18} u_1 + \frac{13}{9} u_2 + \frac{2}{18} u_3 \\ &= \frac{1}{3} u_1 + \frac{13}{9} u_2 + \frac{1}{9} u_3, \end{aligned}$$

$$\text{So } [x]_U = \begin{bmatrix} \frac{1}{3} \\ \frac{13}{9} \\ \frac{1}{9} \end{bmatrix}.$$

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Definition A set $\{u_1, \dots, u_k\}$ in \mathbb{R}^n is an
orthonormal set

if it is an orthogonal set of unit vectors.

The simplest example is the standard basis of \mathbb{R}^n .

You can take any set of nonzero orthogonal vectors,
and normalize them all, to obtain an
orthonormal set.

(They're still pairwise orthogonal after normalizing.)

Suppose $\{u_1, u_2, u_3\}$ is an orthonormal set in \mathbb{R}^3 (11)

and let $U = \begin{bmatrix} u_1 & u_2 & u_3 \end{bmatrix}$. Let's calculate

$$U^T U = \begin{bmatrix} -u_1^T \\ -u_2^T \\ -u_3^T \end{bmatrix} \begin{bmatrix} u_1 & u_2 & u_3 \end{bmatrix} = \begin{bmatrix} u_1 \cdot u_1 & u_1 \cdot u_2 & u_1 \cdot u_3 \\ u_2 \cdot u_1 & u_2 \cdot u_2 & u_2 \cdot u_3 \\ u_3 \cdot u_1 & u_3 \cdot u_2 & u_3 \cdot u_3 \end{bmatrix}$$

$$\stackrel{\text{(by orthogonality)}}{=} \begin{bmatrix} u_1 \cdot u_1 & 0 & 0 \\ 0 & u_2 \cdot u_2 & 0 \\ 0 & 0 & u_3 \cdot u_3 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \begin{array}{l} \text{(because)} \\ \text{(unit vectors!)} \end{array}$$

Thus $U^T U = I_3$, and so $U U^T = I_3$ as well.

Definition a square matrix U with orthonormal columns
is called an orthogonal matrix. (12)

Theorem orthogonal matrices are automatically invertible,
and $U^{-1} = U^T$.

Theorem • an $m \times n$ matrix U has orthonormal columns
if and only if $U^T U = I$.

- in that case, for any vectors x, y
 - $\|Ux\| = \|x\|$
 - $(Ux) \cdot (Uy) = x \cdot y$

Example

Consider the 4×3 matrix

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

It has orthogonal columns,

and $A^T A = \begin{bmatrix} 6 & 0 & 0 \\ 0 & 12 & 0 \\ 0 & 0 & 6 \end{bmatrix}$.

- Are the rows of A orthogonal?
- Obtain B by multiplying each column of A by a scalar, so $B^T B = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 12 & 0 \\ 0 & 0 & 6 \end{bmatrix}$

- the rows are not orthogonal:

$$\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 2 \\ -1 \\ -1 \end{bmatrix} = 2 - 1 - 2 = -1 \neq 0.$$

- we should **normalize** the columns!

$$\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} = 6, \quad \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} = 12, \quad \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix} \cdot \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix} = 6,$$

so $B = \begin{bmatrix} \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{12}} & \frac{2}{\sqrt{6}} \\ \frac{2}{\sqrt{6}} & -\frac{1}{\sqrt{12}} & -\frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{12}} & 0 \\ 0 & \frac{3}{\sqrt{12}} & -\frac{1}{\sqrt{6}} \end{bmatrix}$

will do.

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