

# Orthogonality

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## Orthonormal set of vectors

set of vectors  $\{u_1, \dots, u_k\} \subset \mathbb{R}^n$  is

- ▶ *normalized* if  $\|u_i\| = 1$ ,  $i = 1, \dots, k$   
( $u_i$  are called *unit vectors* or *direction vectors*)
- ▶ *orthogonal* if  $u_i \perp u_j$  for  $i \neq j$
- ▶ *orthonormal* if both

**slang:** we say ' $u_1, \dots, u_k$  are orthonormal vectors' but orthonormality (like independence) is a property of a *set* of vectors, not vectors individually

in terms of  $U = [u_1 \ \cdots \ u_k]$ , orthonormal means

$$U^T U = I_k$$

## Orthonormality

an orthonormal set of vectors is independent

- ▶ to see this, multiply  $Ux = 0$  by  $U^T$
- ▶ hence  $\{u_1, \dots, u_k\}$  is an *orthonormal basis* for

$$\text{span}(u_1, \dots, u_k) = \text{range}(U)$$

- ▶ **warning:** if  $k < n$  then  $UU^T \neq I$  (since its rank is at most  $k$ )  
(more on this matrix later ...)

## Orthonormal basis for $\mathbb{R}^n$

A matrix  $U$  is called *orthogonal* if

$$U \text{ is square and } U^T U = I$$

- ▶ the set of columns  $u_1, \dots, u_n$  is an orthonormal *basis* for  $\mathbb{R}^n$
- ▶ (you'd think such matrices would be called *orthonormal*, not *orthogonal*)
- ▶ it follows that  $U^{-1} = U^T$ , and hence also  $U U^T = I$ , *i.e.*,

$$\sum_{i=1}^n u_i u_i^T = I$$

## Expansion in orthonormal basis

suppose  $U$  is orthogonal, so  $x = UU^T x$ , i.e.,

$$x = \sum_{i=1}^n (u_i^T x) u_i$$

- ▶  $u_i^T x$  is called the *component* of  $x$  in the direction  $u_i$
- ▶  $a = U^T x$  *resolves*  $x$  into the vector of its  $u_i$  components
- ▶  $x = Ua$  *reconstitutes*  $x$  from its  $u_i$  components
- ▶  $x = Ua = \sum_{i=1}^n a_i u_i$  is called the ( $u_i$ -) *expansion* of  $x$

## Complementary subspaces

if  $Q = [Q_1 \quad Q_2]$  and  $Q$  is orthogonal then  $\text{range}(Q_1)$  and  $\text{range}(Q_2)$  are called *complementary subspaces*, because

$$\text{range}(Q_2) = \text{range}(Q_1)^\perp$$

- ▶ they are orthogonal *i.e.*, every vector in the first subspace is orthogonal to every vector in the second subspace
- ▶ every vector in  $\mathbb{R}^m$  can be expressed as a sum of two vectors, one from each subspace
- ▶ each subspace is the *orthogonal complement* of the other

## Complementary subspaces

$$\text{range}(Q_2) = \text{range}(Q_1)^\perp$$

- ▶  $\text{range } Q_2 \subset (\text{range } Q_1)^\perp$  because  $Q_1^\top Q_2 = 0$
- ▶ to show  $\text{range } Q_2 \supset (\text{range } Q_1)^\perp$ , suppose  $x \in (\text{range } Q_1)^\perp$ , then  $Q_1^\top x = 0$ , and since  $x = Q_1 Q_1^\top x + Q_2 Q_2^\top x$  we have  $x = Q_2 Q_2^\top x$  and so  $x \in \text{range } Q_2$

## Geometric interpretation

if  $U$  has orthonormal columns then transformation  $w = Uz$

- ▶ preserves *norm* of vectors, *i.e.*,  $\|Uz\| = \|z\|$
- ▶ preserves *angles* between vectors, *i.e.*,  $\angle(Uz, U\tilde{z}) = \angle(z, \tilde{z})$
- ▶ we say  $U$  is *isometric*, it preserves distances

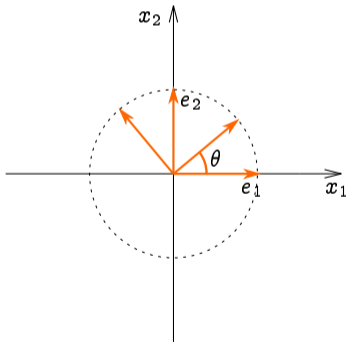


## Example: Rotation

rotation by  $\theta$  in  $\mathbb{R}^2$  is given by

$$y = U_\theta x, \quad U_\theta = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

since  $e_1 \mapsto (\cos \theta, \sin \theta)$ ,  $e_2 \mapsto (-\sin \theta, \cos \theta)$



## Example: Reflection

reflection across line  $x_2 = x_1 \tan(\theta/2)$  is given by

$$y = R_\theta x, \quad R_\theta = \begin{bmatrix} \cos \theta & \sin \theta \\ \sin \theta & -\cos \theta \end{bmatrix}$$

since  $e_1 \rightarrow (\cos \theta, \sin \theta)$ ,  $e_2 \rightarrow (\sin \theta, -\cos \theta)$

