

# Orthogonality

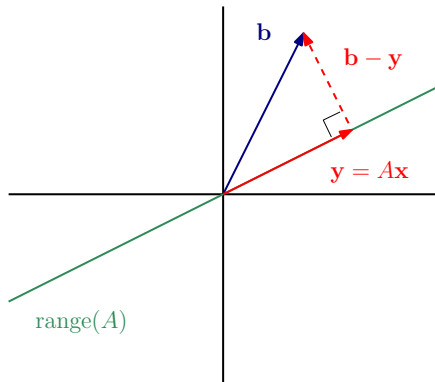
# Orthogonality

# Normal Equation Revisited

Alternate perspective on the “normal equation”:

$$A^T(\mathbf{b} - A\mathbf{x}) = \mathbf{0} \iff \underbrace{\mathbf{z}^T(\mathbf{b} - A\mathbf{x})}_{\text{residual} = \mathbf{r}} = 0 \quad \text{for all } \mathbf{z} \in \mathcal{R}(A),$$

i.e.,  $\mathbf{x}$  solves the normal equation if and only if the residual is orthogonal to the range of  $A$ .



# Orthogonal Vectors

Recall that the angle  $\theta$  between two vectors  $\mathbf{u}, \mathbf{v} \in \mathbb{R}^n$  satisfies

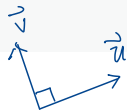
$$\begin{aligned} \vec{u} \cdot \vec{v} &= \|\vec{u}\| \|\vec{v}\| \cos \theta \\ \parallel \\ \vec{u}^T \vec{v} \end{aligned}$$

$$\cos(\theta) = \frac{\mathbf{u}^T \mathbf{v}}{\|\mathbf{u}\|_2 \|\mathbf{v}\|_2}.$$

If  $\theta = \pm \pi/2$ ,

$\cos(\theta) = 0$ , so

$$\vec{u}^T \vec{v} = 0$$



## Definition 1

- Two vectors  $\mathbf{u}, \mathbf{v} \in \mathbb{R}^n$  are **orthogonal** if  $\mathbf{u}^T \mathbf{v} = 0$ .
- Vectors  $\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_k \in \mathbb{R}^n$  are **orthogonal** if  $\mathbf{q}_i^T \mathbf{q}_j = 0$  for all  $i \neq j$ .
- Vectors  $\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_k \in \mathbb{R}^n$  are **orthonormal** if  $\mathbf{q}_i^T \mathbf{q}_j = \delta_{i,j} = \begin{cases} 0, & i \neq j \\ 1, & i = j \end{cases}$

i.e., pairwise orthogonal

**Notation.** (Kronecker delta function)

$$\delta_{i,j} = \begin{cases} 1, & i = j \\ 0, & i \neq j \end{cases}$$

perpendicularity

unit length

$$\vec{q}_i^T \vec{q}_i = 1 = \|\vec{q}_i\|_2^2 \Rightarrow \|\vec{q}_i\|_2 = 1$$

i.e.,  $\vec{q}_i$  is unit.

# Matrices with Orthogonal Columns

Let  $Q = [\mathbf{q}_1 \mid \mathbf{q}_2 \mid \cdots \mid \mathbf{q}_k] \in \mathbb{R}^{n \times k}$ . Note that

Note:  $(i,j)$ -entry of  $Q^T Q = \vec{q}_i^T \vec{q}_j$

$$Q^T Q = \begin{bmatrix} \mathbf{q}_1^T \\ \mathbf{q}_2^T \\ \vdots \\ \mathbf{q}_k^T \end{bmatrix} \begin{bmatrix} \mathbf{q}_1 & \mathbf{q}_2 & \cdots & \mathbf{q}_k \end{bmatrix} = \begin{bmatrix} \mathbf{q}_1^T \mathbf{q}_1 & \mathbf{q}_1^T \mathbf{q}_2 & \cdots & \mathbf{q}_1^T \mathbf{q}_k \\ \mathbf{q}_2^T \mathbf{q}_1 & \mathbf{q}_2^T \mathbf{q}_2 & \cdots & \mathbf{q}_2^T \mathbf{q}_k \\ \vdots & \vdots & \ddots & \vdots \\ \mathbf{q}_k^T \mathbf{q}_1 & \mathbf{q}_k^T \mathbf{q}_2 & \cdots & \mathbf{q}_k^T \mathbf{q}_k \end{bmatrix}.$$

Therefore,

$\vec{q}_i^T \vec{q}_j = 0 \Leftrightarrow$  all off-diagonal terms are 0's.

- $\mathbf{q}_1, \dots, \mathbf{q}_k$  are orthogonal.  $\Leftrightarrow Q^T Q$  is a  $k \times k$  diagonal matrix.
- $\mathbf{q}_1, \dots, \mathbf{q}_k$  are orthonormal.  $\Leftrightarrow Q^T Q$  is the  $k \times k$  identity matrix.

because  $\vec{q}_i^T \vec{q}_i = 1$ .

# Matrices with Orthonormal Columns

$$\vec{q}_i^T \vec{q}_j = \delta_{ij} = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$$

## Theorem 2

Let  $Q = [\mathbf{q}_1 \mid \mathbf{q}_2 \mid \cdots \mid \mathbf{q}_k] \in \mathbb{R}^{n \times k}$  and suppose that  $\mathbf{q}_1, \dots, \mathbf{q}_k$  are orthonormal. Then

- 1  $Q^T Q = I \in \mathbb{R}^{k \times k}$ ;
- 2  $\|Q\mathbf{x}\|_2 = \|\mathbf{x}\|_2$  for all  $\mathbf{x} \in \mathbb{R}^k$ ; (2-norm preservation)
- 3  $\|Q\|_2 = 1$ . (exercise; recall the def'n of matrix p-norm & use 2)

Proof of 2

$$\|Q\vec{x}\|_2^2 = (Q\vec{x})^T (Q\vec{x}) = \vec{x}^T \underbrace{Q^T Q}_{=I}_{\text{by 1}} \vec{x} = \vec{x}^T \vec{x} = \|\vec{x}\|_2^2$$

# Orthogonal Matrices

## Definition 3

We say that  $Q \in \mathbb{R}^{n \times n}$  is an **orthogonal matrix** if  $Q^T Q = I \in \mathbb{R}^{n \times n}$ .

Square matrix

The columns of  $Q$  are orthonormal.

- A square matrix with orthogonal columns is not, in general, an orthogonal matrix!

e.g.

$$\cdot I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

- $P$  permutation matrix.

$\vec{r}_1$   
↓

$\vec{r}_2$   
↓

$$\cdot R_\theta = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

$$\vec{r}_1^T \vec{r}_2 = \cos \theta (-\sin \theta)$$

$$+ \sin \theta \cdot \cos \theta = 0$$

$$\|\vec{r}_1\|_2^2 = \cos^2 \theta + \sin^2 \theta = 1$$

$$\|\vec{r}_2\|_2^2 = \sin^2 \theta + \cos^2 \theta = 1$$

# Properties of Orthogonal Matrices

$$Q^T Q = I$$

## Theorem 4

Let  $Q \in \mathbb{R}^{n \times n}$  be orthogonal. Then

- 1  $Q^{-1} = Q^T$ ;
- 2  $Q^T$  is also an orthogonal matrix;
- 3  $\kappa_2(Q) = 1$ ; the best that we can hope for!
- 4 For any  $A \in \mathbb{R}^{n \times n}$ ,  $\|AQ\|_2 = \|A\|_2$ ;
- 5 if  $P \in \mathbb{R}^{n \times n}$  is another orthogonal matrix, then  $PQ$  is also orthogonal.

$$(PQ)^T(PQ) = Q^T \underbrace{P^T P}_I Q = Q^T Q = I$$

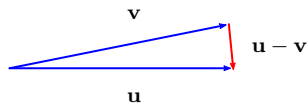


# Why Do We Like Orthogonal Vectors?

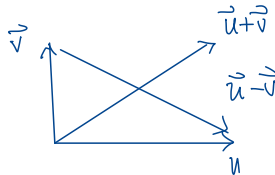
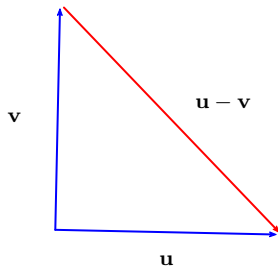
- If  $\mathbf{u}$  and  $\mathbf{v}$  are orthogonal, then

$$\|\mathbf{u} \pm \mathbf{v}\|_2^2 = \|\vec{u}\|_2^2 + \|\vec{v}\|_2^2 \pm 2\vec{u}^T\vec{v} \quad (\text{Pythagorean theorem})$$

- Without orthogonality, it is possible that  $\|\mathbf{u} - \mathbf{v}\|_2$  is much smaller than  $\|\mathbf{u}\|_2$  and  $\|\mathbf{v}\|_2$ .
- The addition and subtraction of orthogonal vectors are guaranteed to be well-conditioned.



Catastrophic cancellation.



# Appendix: Projection and Reflection

# Projection and Reflection Operators

Let  $\mathbf{u}, \mathbf{v} \in \mathbb{R}^m$  be nonzero vectors.

- Projection of  $\mathbf{u}$  onto  $\langle \mathbf{v} \rangle = \text{span}(\mathbf{v})$ :

$$\frac{\mathbf{v}^T \mathbf{u}}{\mathbf{v}^T \mathbf{v}} \mathbf{v} = \underbrace{\left( \frac{\mathbf{v} \mathbf{v}^T}{\mathbf{v}^T \mathbf{v}} \right)}_{=: P} \mathbf{u} =: P \mathbf{u}.$$

- Projection of  $\mathbf{u}$  onto  $\langle \mathbf{v} \rangle^\perp$ , the orthogonal complement of  $\langle \mathbf{v} \rangle$ :

$$\mathbf{u} - \frac{\mathbf{v}^T \mathbf{u}}{\mathbf{v}^T \mathbf{v}} \mathbf{v} = \left( I - \frac{\mathbf{v} \mathbf{v}^T}{\mathbf{v}^T \mathbf{v}} \right) \mathbf{u} =: (I - P) \mathbf{u}.$$

- Reflection of  $\mathbf{u}$  across  $\langle \mathbf{v} \rangle^\perp$ :

$$\mathbf{u} - 2 \frac{\mathbf{v}^T \mathbf{u}}{\mathbf{v}^T \mathbf{v}} \mathbf{v} = \left( I - 2 \frac{\mathbf{v} \mathbf{v}^T}{\mathbf{v}^T \mathbf{v}} \right) \mathbf{u} =: (I - 2P) \mathbf{u}.$$

## Projection and Reflection Operators (cont')

**Summary:** for given  $\mathbf{v} \in \mathbb{R}^m$ , a nonzero vector, let

$$P = \frac{\mathbf{v}\mathbf{v}^T}{\mathbf{v}^T\mathbf{v}} \in \mathbb{R}^{m \times m}.$$

Then the following matrices carry out geometric transformations

- Projection onto  $\langle \mathbf{v} \rangle$ :  $P$
- Projection onto  $\langle \mathbf{v} \rangle^\perp$ :  $I - P$
- Reflection across  $\langle \mathbf{v} \rangle^\perp$ :  $I - 2P$

**Note.** If  $\mathbf{v}$  were a unit vector, the definition of  $P$  simplifies to  $P = \mathbf{v}\mathbf{v}^T$ .