# Orthogonality

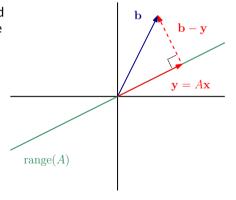
# Orthogonality

#### Normal Equation Revisited

Alternate perspective on the "normal equation":

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

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

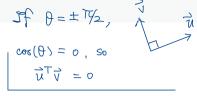


#### **Orthogonal Vectors**

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

$$\vec{\mathcal{N}} \cdot \vec{\mathcal{V}} = \|\vec{\mathcal{M}}\| \|\vec{\mathcal{V}}\| \cos \theta$$

$$\lim_{n \to \infty} \cos(\theta) = \frac{\mathbf{u}^{\mathrm{T}} \mathbf{v}}{\|\mathbf{u}\|_{2} \|\mathbf{v}\|_{2}}.$$



#### **Definition 1**

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- 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  $\mathbf{q}_i^{\mathrm{T}} \mathbf{q}_j = 0$  for all  $i \neq j$ .

• Vectors 
$$\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_k \in \mathbb{R}^n$$
 are orthonormal if  $\mathbf{q}_i^T \mathbf{q}_j = \delta_{i,j} = \emptyset$ ,  $\hat{\iota} \neq \hat{\jmath}$ 

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

## Matrices with Orthogonal Columns

$$\frac{N_{\text{ate}}}{\sqrt{N_{\text{o}}}}$$
:  $(i,j)$ -entry =  $\vec{q}_i$  $\vec{q}_i$ 

Let 
$$Q = \left[ \begin{array}{c|c} \mathbf{q}_1 & \mathbf{q}_2 & \cdots & \mathbf{q}_k \end{array} \right] \in \mathbb{R}^{n \times k}$$
 . Note that

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

$$\overrightarrow{q_i} \overrightarrow{q_j} = 0 \iff \text{all off-diagonal}$$
 Therefore, 
$$\overrightarrow{q_1, \dots, q_k} \text{ are orthogonal.} \iff Q^TQ \text{ is a } k \times k \text{ diagonal matrix.}$$

- $\mathbf{q}_1, \dots, \mathbf{q}_k$  are orthonormal.  $\iff$   $Q^TQ$  is the  $k \times k$  identity matrix.

## Matrices with Orthonormal Columns

$$\vec{q}_i \vec{T} \vec{q}_j = \delta_{i,j} = \int_{-\infty}^{\infty} 0 \quad \text{if } i \neq j$$

#### 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  $\mathbf{q}_1, \ldots, \mathbf{q}_k$  are orthonormal. Then

- 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 defining matrix p-norm & use 2)

Proof of 
$$Q$$

$$||Q\overrightarrow{x}||_{2}^{2} = (Q\overrightarrow{x})^{T}(Q\overrightarrow{x}) = \overrightarrow{x}^{T}Q^{T}Q\overrightarrow{x} = \overrightarrow{x}^{T}\overrightarrow{x} = ||\overrightarrow{x}||_{2}^{2}$$

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### **Orthogonal Matrices**

#### **Definition 3**

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

Equare matrix The columns of a are orthonormal.

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

$$R_0 = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

### **Properties of Orthogonal Matrices**

#### Theorem 4

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

- $\mathbf{0} \ Q^{-1} = Q^{\mathrm{T}};$
- Q Q is also an orthogonal matrix;
- 3  $\kappa_2(Q)=1;$  the best that we can hope for !
- **4** For any  $A \in \mathbb{R}^{n \times n}$ ,  $||AQ||_2 = ||A||_2$ ;
- **6** if  $P \in \mathbb{R}^{n \times n}$  is another orthogonal matrix, then PQ is also orthogonal.

$$(PQ)^T(PQ) = Q^T P^T PQ = Q^T Q = I$$

### Why Do We Like Orthogonal Vectors?

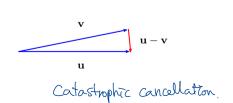
If u and v are orthogonal, then

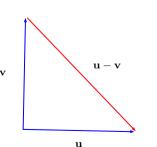
$$\|\mathbf{u} \pm \mathbf{v}\|_{2}^{2} = \|\vec{\mathcal{U}}\|_{2}^{2} + \|\vec{\mathcal{V}}\|_{2}^{2} \pm 2\vec{\mathcal{U}}$$
 (Pythagorean theorem)

• 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.





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# Appendix: Projection and Reflection

(Review of Linear Afgebora)

Given 
$$\vec{u}$$
,  $\vec{v} \in \mathbb{R}^{n}$ 

Today:  $||\cdot||$  for  $||\cdot||_{2}$ 

$$P = \frac{\vec{v} \vec{v}^{T}}{\vec{v}^{T} \vec{v}^{T}}$$

$$P = ||\vec{p}|| (\text{unit vector} |\vec{v}| + \text{div.} \sqrt{r} \vec{v})$$

$$= ||\vec{p}|| \frac{\vec{v}}{||\vec{v}||}$$

$$= ||\vec{v}|| |\cos \theta||\vec{v}||_{2}$$

$$|\vec{v}| = \text{span } |\vec{v}|$$
Note that
$$|\vec{p} + \vec{q}| = |\vec{v}|$$

$$|\vec{p}| = |\vec{v}| - |\vec{p}|$$

$$|\vec{v}| = |\vec{v}| - |\vec{p}|$$

$$|\vec{v}| = |\vec{v}| - |\vec{p}|$$

$$|\vec{v}| = |\vec{v}| - |\vec{v}|$$

$$|\vec{r}| = |\vec{q}| + (-\vec{p})$$

$$= |\vec{T} - 2\vec{p}| = |\vec{T} -$$

$$\langle \vec{V} \rangle^{+} = \text{orthogonal complement}$$
of  $\langle \vec{V} \rangle$ .

### **Projection and Reflection Operators**

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

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

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

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

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

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

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

#### Projection and Reflection Operators (cont')

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

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

Then the following matrices carry out geometric transformations

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

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