

# Recitation 5

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Fall 2020

# Orthogonal matrices

## Definition (Orthogonal matrix)

Given a dot product  $\langle \cdot, \cdot \rangle$ , an orthogonal matrix is a real square matrix whose columns are *orthonormal* vectors.

Recall:

- ❖  $Q$  is an orthonormal matrix iff its inverse is  $Q^T$ .
- ❖  $\langle Qx, Qy \rangle = \langle x, y \rangle$  for all  $x, y$  with the appropriate dimensions and  $Q$  orthogonal.
- ❖ Show that  $\|Qx\| = \|x\|$  for all  $x$  and  $Q$  orthogonal.

# Questions: Gram-Schmidt and QR

1. Let  $A \in \mathbb{R}^{n \times n}$  have linearly independent columns. Show that there is a matrix  $Q \in \mathbb{R}^{n \times n}$  and  $R \in \mathbb{R}^{n \times n}$  s.t that  $A = QR$ , where  $Q$  has orthonormal columns and  $R$  is upper triangular. (Hint: Recall the “linear combination of columns interpretation of matrix multiplication”). What if the columns are not linearly independent?

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# Eigenvalues and eigenvectors

## Definition

Let  $A \in \mathbb{R}^{n \times n}$ . A **non-zero** vector  $v \in \mathbb{R}^n$  is said to be an *eigenvector* of  $A$  if there exists  $\lambda \in \mathbb{R}$  such that

$$Av = \lambda v.$$

The scalar  $\lambda$  is called the *eigenvalue* (of  $A$ ) associated to  $v$ . The set

$$E_\lambda(A) = \{x \in \mathbb{R}^n \mid Ax = \lambda x\} = \text{Ker}(A - \lambda \text{Id})$$

is called the *eigenspace* of  $A$  associated to  $\lambda$ . The dimension of  $E_\lambda(A)$  is called the multiplicity of the eigenvalue  $\lambda$ .

# Eigenvalues and eigenvectors

Recall:

- ❖ If a matrix  $A \in \mathbb{R}^{n \times n}$  has eigenvalues  $\lambda_1 < \dots < \lambda_k$  with eigenvectors  $v_1, \dots, v_k$  resp., then  $v_1, \dots, v_k$  are linearly independent.  $\implies A$  has at most  $n$  different eigenvalues.
- ❖ More strongly, if  $A \in \mathbb{R}^{n \times n}$  has eigenvalues  $\lambda_1 < \dots < \lambda_k$

$$\sum_{i=1}^k \dim(E_{\lambda_i}(A)) \leq n$$

Note: To compute eigenvalues and eigenvectors using determinants and characteristic polynomials, see Léo's video. Recommended but optional and not covered in this recitation.

# Questions: Eigendecomposition

1. Let  $V = \begin{bmatrix} v_1 & \cdots & v_n \end{bmatrix} \in \mathbb{R}^{n \times n}$ . Show that a matrix  $A \in \mathbb{R}^{n \times n}$  has eigenvalues  $\lambda_1, \dots, \lambda_n$  with linearly independent eigenvectors  $v_1, \dots, v_n$  iff  $A = V \operatorname{diag}((\lambda_i)_{i=1}^n) V^{-1}$ .

## Definition

We say that  $A \in \mathbb{R}^{n \times n}$  is a *diagonalizable* matrix if it has eigenvalues  $\lambda_1, \dots, \lambda_n$  with linearly independent eigenvectors  $v_1, \dots, v_n$ .

2. Show that if  $A \in \mathbb{R}^{n \times n}$  has eigenvalues  $\lambda_1 < \cdots < \lambda_n$ ,  $A$  is diagonalizable.
3. Write the expression of a matrix in  $\mathbb{R}^{2 \times 2}$  for which  $\begin{bmatrix} 2 \\ -1 \end{bmatrix}$  is an eigenvector of eigenvalue 2 and  $\begin{bmatrix} 1 \\ 3 \end{bmatrix}$  is an eigenvector of eigenvalue  $-1$ .



# Questions: Eigendecomposition

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3. Write the expression of a matrix in  $\mathbb{R}^{2 \times 2}$  for which  $[2, -1]$  is an eigenvector of eigenvalue 2 and  $[1, 3]$  is an eigenvector of eigenvalue  $-1$ .

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# Questions: Eigenvalues & trace

1. Show that the trace is invariant by change of basis, i.e. if  $X \in \mathbb{R}^{n \times n}$  is invertible and  $A \in \mathbb{R}^{n \times n}$ ,  $\text{tr}(A) = \text{tr}(XAX^{-1})$ . (Hint:  $\text{tr}(BC) = \text{tr}(CB)$ ).
2. Show that if  $A \in \mathbb{R}^{n \times n}$  is diagonalizable and has eigenvalues  $\lambda_1, \dots, \lambda_n$ , then  $\text{tr}(A) = \sum_{i=1}^n \lambda_i$ .

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2. Show that if  $A \in \mathbb{R}^{n \times n}$  is diagonalizable and has eigenvalues  $\lambda_1, \dots, \lambda_n$ , then  $\text{tr}(A) = \sum_{i=1}^n \lambda_i$ .

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# Questions: No (real) eigenvalues

Some matrices do not admit (real) eigenvalues and eigenvectors.

1. Show that if  $\theta \in [0, 2\pi)$ ,

$$\begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}$$

does not admit real eigenvalues and eigenvectors in general.

2. Find the matrices, and the real eigenvalues and eigenvectors for the values of  $\theta$  for which they exist.

# Questions: No (real) eigenvalues

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admit real eigenvalues and eigenvectors?

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