## COS 302 Precept 4

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Princeton University



Geometric interpretation



Geometric interpretation

## Definition

Consider a square matrix  $D \in \mathbb{R}^{n \times n}$ . D is a diagonal matrix if all  $D_{ij} = 0$  if  $i \neq j$ , i.e., it is of the form:

$$D = \begin{bmatrix} d_1 & 0 \\ & \ddots & \\ 0 & & d_n \end{bmatrix}$$

## **Properties of Diagonal Matrix**

- Diagonal matrices allow fast computation of determinants, powers, and inverses
- The determinant of a diagonal matrix is the product of its diagonal entries
- A matrix power **D**<sup>k</sup> is given by each diagonal element raised to the power k
- *D*<sup>-1</sup> is the reciprocal of its diagonal elements if all of them are nonzero.

## **Diagonalizable Matrix**

## Definition

A matrix  $A \in \mathbb{R}^{n \times n}$  is diagonalizable if there exists an invertible matrix  $P \in \mathbb{R}^{n \times n}$  such that  $D = P^{-1}AP$ , where D is a diagonal matrix.

#### Theorem

A matrix  $\boldsymbol{A} \in \mathbb{R}^{n \times n}$  can be factored into

## $A = PDP^{-1}$

where  $P \in \mathbb{R}^{n \times n}$  and D is a diagonal matrix whose diagonal entries are the eigenvalues of A, if and only if the eigenvectors of A form a basis of  $\mathbb{R}^n$ . This factorization is known as the eigendecomposition of A.

## **Proof of Eigendecomposition Theorem**

#### Proof

Let  $A \in \mathbb{R}^{n \times n}$ , let  $\lambda_1, \lambda_2, \dots, \lambda_n$  be a set of scalars, and  $p_1, p_2, \dots, p_n$  be a set of vectors in  $\mathbb{R}^n$ . We define  $P := [p_1, p_2, \dots, p_n]$ . That is,  $p_i$ 's are the columns of P.Let  $D \in \mathbb{R}^{n \times n}$  be a diagonal matrix whose diagonal entries are  $\lambda_1, \lambda_2, \dots, \lambda_n$ . That is,

$$D = \begin{bmatrix} \lambda_1 & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$$

## **Proof of Eigendecomposition Theorem**

#### **Proof Continued**

$$\boldsymbol{AP} = [\boldsymbol{Ap_1}, \boldsymbol{Ap_2}, \cdots, \boldsymbol{Ap_n}] \tag{1}$$

$$\boldsymbol{P}\boldsymbol{D} = [\lambda_1 \boldsymbol{p}_1, \lambda_2 \boldsymbol{p}_2, \cdots, \lambda_n \boldsymbol{p}_n]$$
(2)

The above two equations imply that:  $Ap_i = \lambda_1 p_i$  $\forall 1 \le i \le n$ . That is, the  $p_i$ 's are the eigenvectors of A and  $\lambda_i$ 's are the corresponding eigenvalues.

## **Proof of Eigendecomposition Theorem**

#### **Proof Continued**

In the definition of diagonalizability, we require that P is invertible, which implies that the columns of P are linearly independent of each other. Therefore, the columns of P, which are also the eigenvectors of A, form a basis of  $\mathbb{R}^n$ .

## Symmetric Square Matrix

#### Theorem

# Symmetric Square Matrices are always diagonalizable.



## Geometric interpretation

## Geometric interpretation of eigen-decomposition



**Figure 1:** Intuition behind the eigendecomposition as sequential transformations.

1.  $P^{-1}$  performs a basis change (here drawn in  $\mathbb{R}^2$ and depicted as a rotation-like operation), mapping the eigenvectors into the standard basis.

• 
$$P\tilde{x}_1 = x \iff \tilde{x}_1 = P^{-1}x$$

2. *D* performs a scaling along the remapped orthogonal eigenvectors, depicted here by a circle being stretched to an ellipse

• 
$$\tilde{x}_2 = D\tilde{x}_1$$

## Geometric interpretation of eigen-decomposition

3. *P* undoes the basis change (depicted as a reverse rotation) and restores the original coordinate frame.

• 
$$\tilde{x} = P\tilde{x}_2$$



## Geometric interpretation

Assume a matrix has eigen-decomposition:  $A = PDP^{-1}$ .

- Calculating matrix power:  $A^{k} = (PDP^{-1})^{k} = PD^{k}P^{-1}$
- Calculating determinant:  $det(A) = det(PDP^{-1})$   $= det(P) det(D) det(P^{-1})$   $= det(D) det(P) det(P^{-1})$   $= det(D) det(PP^{-1})$  = det(D)