

# Lecture 38: Eigenvectors and Eigenvalues of Inverse Matrices and Matrix Transposes

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MATH3200: Applied Linear Algebra

# An example

In this lecture we will explore how the eigenvalues and eigenvectors of a square matrix  $\mathbf{A}$  are related to the eigenvalues and eigenvectors of  $\mathbf{A}^{-1}$  and  $\mathbf{A}^T$ .

$$\text{Let } \mathbf{A} = \begin{bmatrix} 1 & 3 \\ 1 & -1 \end{bmatrix} \quad \text{and let} \quad \mathbf{B} = \begin{bmatrix} 0.25 & 0.75 \\ 0.25 & -0.25 \end{bmatrix}$$

$$\text{Then } \mathbf{AB} = \begin{bmatrix} 1 & 3 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 0.25 & 0.75 \\ 0.25 & -0.25 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \text{Thus } \mathbf{B} = \mathbf{A}^{-1}.$$

In Lecture 36 we had explored the matrix  $\mathbf{A}$  and found:

- $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda_1 = -2$ .
- $\begin{bmatrix} 3 \\ 1 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda_2 = 2$ .

## An example, continued

$$\text{Let } \mathbf{A} = \begin{bmatrix} 1 & 3 \\ 1 & -1 \end{bmatrix} \quad \text{Then } \mathbf{A}^{-1} = \begin{bmatrix} 0.25 & 0.75 \\ 0.25 & -0.25 \end{bmatrix}$$

$$\vec{x}_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \text{ is an eigenvector of } \mathbf{A} \text{ with eigenvalue } \lambda_1 = -2.$$

**Question L38.1:** Is  $\vec{x}_1$  also an eigenvector of  $\mathbf{A}^{-1}$ ?  
If so, what is its eigenvalue?

$$\vec{x}_2 = \begin{bmatrix} 3 \\ 1 \end{bmatrix} \text{ is an eigenvector of } \mathbf{A} \text{ with eigenvalue } \lambda_2 = 2.$$

**Question L38.2:** Is  $\vec{x}_2$  also an eigenvector of  $\mathbf{A}^{-1}$ ?  
If so, what is its eigenvalue?

# An example: Eigenvalues and eigenvectors of $\mathbf{A}^{-1}$

Let  $\mathbf{A} = \begin{bmatrix} 1 & 3 \\ 1 & -1 \end{bmatrix}$  Then  $\mathbf{A}^{-1} = \begin{bmatrix} 0.25 & 0.75 \\ 0.25 & -0.25 \end{bmatrix}$

- $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda_1 = -2$ .
- $\begin{bmatrix} 0.25 & 0.75 \\ 0.25 & -0.25 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} -0.5 \\ 0.5 \end{bmatrix} = \frac{1}{-2} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$
- $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}^{-1}$  with eigenvalue  $\frac{1}{\lambda_1} = \frac{1}{-2}$ .
- $\begin{bmatrix} 3 \\ 1 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda_2 = 2$ .
- $\begin{bmatrix} 0.25 & 0.75 \\ 0.25 & -0.25 \end{bmatrix} \begin{bmatrix} 3 \\ 1 \end{bmatrix} = \begin{bmatrix} 1.5 \\ 0.5 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 3 \\ 1 \end{bmatrix}$
- $\begin{bmatrix} 3 \\ 1 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}^{-1}$  with eigenvalue  $\frac{1}{\lambda_2} = \frac{1}{2}$ .

# Eigenvalues and eigenvectors of inverse matrices

This example illustrates the following general result:

## Theorem

*Let  $\mathbf{A}$  be an invertible matrix.*

*Let  $\vec{x}$  be an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda$ .*

*Then  $\vec{x}$  is an eigenvector of  $\mathbf{A}^{-1}$  with eigenvalue  $\frac{1}{\lambda}$ .*

In other words,  $\mathbf{A}$  and  $\mathbf{A}^{-1}$  have the same eigenvectors, and the eigenvalues of  $\mathbf{A}^{-1}$  are the reciprocals of the eigenvalues of  $\mathbf{A}$ .

## How about the eigenvectors of the transpose?

$$\text{Let } \mathbf{A} = \begin{bmatrix} 8 & 3 \\ -6 & -1 \end{bmatrix} \quad \text{Then } \mathbf{A}^T = \begin{bmatrix} 8 & -6 \\ 3 & -1 \end{bmatrix}$$

in Module 67 we found that:

- $\vec{x}_1 = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda_1 = 2$ .
- $\vec{x}_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda_2 = 5$ .

**Question L38.3:** Is the vector  $\vec{x}_2$  above an eigenvector of  $\mathbf{A}^T$ ?  
If so, what is its eigenvalue?

$$\mathbf{A}^T \vec{x}_2 = \begin{bmatrix} 8 & -6 \\ 3 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 14 \\ 4 \end{bmatrix} \neq \lambda \begin{bmatrix} 1 \\ -1 \end{bmatrix}.$$

Thus  $\vec{x}_2$  is not an eigenvector of  $\mathbf{A}$ . Similarly, neither is  $\vec{x}_1$ .

## So how about the eigenvectors of the transpose?

$$\text{Let } \mathbf{A} = \begin{bmatrix} 8 & 3 \\ -6 & -1 \end{bmatrix} \quad \text{Then } \mathbf{A}^T = \begin{bmatrix} 8 & -6 \\ 3 & -1 \end{bmatrix}$$

in Module 67 and Lecture 36 we found that:

- $\begin{bmatrix} 1 \\ -2 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda_1 = 2$ .
- $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda_2 = 5$ .
- $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}^T$  with eigenvalue  $\lambda_2 = 2$ .
- $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}^T$  with eigenvalue  $\lambda_1 = 5$ .

Thus in this example, the eigenvalues of  $\mathbf{A}$  and  $\mathbf{A}^T$  are the same, but the eigenvectors of these matrices are different.

# A theorem about the eigenvalues of the transpose

The preceding example illustrates the following general theorem.

## Theorem

*Let  $\mathbf{A}$  be a square matrix, and let  $\lambda$  be an eigenvalue of  $\mathbf{A}$ .  
Then  $\lambda$  is an eigenvalue of  $\mathbf{A}^T$ .*

Thus a square matrix  $\mathbf{A}$  and its transpose  $\mathbf{A}^T$  always have the same eigenvalues, but not necessarily the same eigenvectors.

In fact, there is no obvious relationship between the eigenvectors of  $\mathbf{A}$  and  $\mathbf{A}^T$ .

However, it turns out that the transposes of the eigenvectors of  $\mathbf{A}$  will always be so-called *left eigenvectors* of  $\mathbf{A}^T$  and *vice versa*.



## (Right) eigenvectors and left eigenvectors

Recall that a *column* vector  $\vec{x} \neq \vec{0}$  is an *eigenvector* with eigenvalue  $\lambda$  of a square matrix  $\mathbf{A}$  if  $\mathbf{A}\vec{x} = \lambda\vec{x}$ .

Eigenvectors are sometimes also called a *right eigenvectors*.

### Definition

A *row* vector  $\vec{x} \neq \vec{0}$  is a *left eigenvector* with eigenvalue  $\lambda$  of a square matrix  $\mathbf{A}$  if  $\vec{x}\mathbf{A} = \lambda\vec{x}$ .

For example,  $[1, -1] \begin{bmatrix} 1 & 4 \\ 2 & 3 \end{bmatrix} = [-1, 1] = (-1)[1, -1]$ .

Thus  $[1, -1]$  is a left eigenvector of  $\mathbf{A} = \begin{bmatrix} 1 & 4 \\ 2 & 3 \end{bmatrix}$

with eigenvalue  $\lambda = -1$ .

# A theorem about left eigenvectors of $\mathbf{A}^T$

Now suppose  $\vec{x}$  is an eigenvector with eigenvalue  $\lambda$  of  $\mathbf{A}$ .

This means that  $\vec{x} \neq \vec{0}$  and  $\mathbf{A}\vec{x} = \lambda\vec{x}$ .

Then  $\vec{x}^T \neq \vec{0}$  and  $(\mathbf{A}\vec{x})^T = \vec{x}^T \mathbf{A}^T = (\lambda\vec{x})^T = \lambda\vec{x}^T$ .

Thus  $\vec{x}^T$  is a left eigenvector of  $\mathbf{A}^T$  with the same eigenvalue  $\lambda$ .

This observation proves the following result:

## Theorem

*Let  $\mathbf{A}$  be a square matrix, and let  $\vec{x}$  be an eigenvector with eigenvalue  $\lambda$  of  $\mathbf{A}$ .*

*Then  $\vec{x}^T$  is a left eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda$ .*

*Similarly, if  $\vec{y}$  is a left eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda$ , then  $\vec{y}^T$  is an eigenvector with eigenvalue  $\lambda$  of  $\mathbf{A}^T$ .*

## How about left eigenvectors of the transpose?

$$\text{Let } \mathbf{A} = \begin{bmatrix} 8 & 3 \\ -6 & -1 \end{bmatrix} \quad \text{Then } \mathbf{A}^T = \begin{bmatrix} 8 & -6 \\ 3 & -1 \end{bmatrix}$$

in Module 67 we found that:

- $\vec{x}_1 = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda_1 = 2$ .
- $\vec{x}_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$  is an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda_2 = 5$ .

**Question L38.4:** Find a left eigenvector of  $\mathbf{A}^T$  with eigenvalue  $\lambda = 2$ .

$$\vec{x}_1^T \mathbf{A}^T = [1, -2] \begin{bmatrix} 8 & -6 \\ 3 & -1 \end{bmatrix} = [2, -4] = 2[1, -2].$$

Thus  $\vec{x}_1^T = [1, -2]$  is a left eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda = 2$ .

# Take-home message

## Theorem

*Let  $\mathbf{A}$  be an invertible matrix, and let  $\vec{x}$  be an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda$ .*

*Then  $\vec{x}$  is an eigenvector of  $\mathbf{A}^{-1}$  with eigenvalue  $\frac{1}{\lambda}$ .*

In other words,  $\mathbf{A}$  and  $\mathbf{A}^{-1}$  have the same eigenvectors, and the eigenvalues of  $\mathbf{A}^{-1}$  are the reciprocals of the eigenvalues of  $\mathbf{A}$ .

## Theorem

*Let  $\mathbf{A}$  be a square matrix. Then  $\mathbf{A}$  and  $\mathbf{A}^T$  have the same eigenvalues. However,  $\mathbf{A}$  and  $\mathbf{A}^T$  usually have different eigenvectors.*

*But if  $\vec{x}$  is an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda$ , then  $\vec{x}^T$  is a **left eigenvector** with eigenvalue  $\lambda$  of  $\mathbf{A}^T$ , which means that  $\vec{x}^T \mathbf{A}^T = \lambda \vec{x}^T$ .*

*Similarly, if  $\vec{y}$  is a left eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda$ , then  $\vec{y}^T$  is an eigenvector with eigenvalue  $\lambda$  of  $\mathbf{A}^T$ .*