

# Algebraic and geometric multiplicity of eigenvalues

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The algebraic multiplicity of an eigenvalue is the number of times it appears as a root of the characteristic polynomial (i.e., the polynomial whose roots are the eigenvalues of a matrix).

The geometric multiplicity of an eigenvalue is the dimension of the linear space of its associated eigenvectors (i.e., its eigenspace).

In this lecture we provide rigorous definitions of the two concepts of algebraic and geometric multiplicity and we prove some useful facts about them.

Looking for a **geometric multiplicity calculator** or a **step-by-step tutorial** on how to calculate the geometric multiplicity? Follow [this link](#).



## Algebraic multiplicity

Let us start with a definition.

**Definition** Let  $A$  be a  $K \times K$  matrix. Denote by  $\lambda_1, \dots, \lambda_K$  the  $K$  possibly repeated eigenvalues of  $A$ , which solve the characteristic equation

$$\det(\lambda I - A) = (\lambda - \lambda_1) \cdot \dots \cdot (\lambda - \lambda_K) = 0$$

We say that an eigenvalue  $\lambda_k$  has algebraic multiplicity  $\mu(\lambda_k) \in \mathbb{N}$  if and only if there are no more and no less than  $\mu(\lambda_k)$  solutions of the characteristic equation equal to  $\lambda_k$ .

Let us see some examples.

**Example** Consider the  $2 \times 2$  matrix

$$A = \begin{bmatrix} 4 & 2 \\ 1 & 2 \end{bmatrix}$$

The characteristic polynomial is

$$\begin{aligned} f(\lambda) &= \det \left( \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \begin{bmatrix} 4 & 2 \\ 1 & 2 \end{bmatrix} \right) \\ &= \det \left( \begin{bmatrix} \lambda - 4 & -2 \\ -1 & \lambda - 2 \end{bmatrix} \right) \\ &= (\lambda - 4) \cdot (\lambda - 2) - (-2) \cdot (-1) \\ &= \lambda^2 - 2\lambda - 4\lambda + 8 - 2 \\ &= \lambda^2 - 6\lambda + 6 \end{aligned}$$

The roots of the polynomial, that is, the solutions of  $f(\lambda) = 0$  are

$$\begin{aligned} \lambda_1 &= 3 + \sqrt{3} \\ \lambda_2 &= 3 - \sqrt{3} \end{aligned}$$

Thus,  $A$  has two distinct eigenvalues. Their algebraic multiplicities are

$$\begin{aligned} \mu(\lambda_1) &= 1 \\ \mu(\lambda_2) &= 1 \end{aligned}$$

because they are not repeated.

**Example** Define the  $2 \times 2$  matrix

$$A = \begin{bmatrix} 1 & 0 \\ 2 & 1 \end{bmatrix}$$

Its characteristic polynomial is

$$\begin{aligned}
 f(\lambda) &= \det \left( \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \begin{bmatrix} 1 & 0 \\ 2 & 1 \end{bmatrix} \right) \\
 &= \det \left( \begin{bmatrix} \lambda - 1 & 0 \\ -2 & \lambda - 1 \end{bmatrix} \right) \\
 &= (\lambda - 1) \cdot (\lambda - 1) - 0 \cdot (-2) \\
 &= (\lambda - 1) \cdot (\lambda - 1)
 \end{aligned}$$

The roots of the polynomial, that is, the solutions of  $f(\lambda) = 0$  are

$$\lambda_1 = 1$$

$$\lambda_2 = 1$$

Thus,  $A$  has one repeated eigenvalue whose algebraic multiplicity is

$$\mu(\lambda_1) = \mu(\lambda_2) = 2$$

## Geometric multiplicity

Recall that each eigenvalue is associated to a **linear space** of eigenvectors, called **eigenspace**.

**Definition** Let  $A$  be a  $K \times K$  matrix. Let  $\lambda_k$  be one of the eigenvalues of  $A$  and denote its associated eigenspace by  $E_k$ . The **dimension** of  $E_k$  is called the geometric multiplicity of the eigenvalue  $\lambda_k$ .

Let's now make some examples.

**Definition** Consider the  $2 \times 2$  matrix

$$A = \begin{bmatrix} 2 & 0 \\ 1 & 1 \end{bmatrix}$$

The characteristic polynomial is

$$\begin{aligned}
 f(\lambda) &= \det \left( \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \begin{bmatrix} 2 & 0 \\ 1 & 1 \end{bmatrix} \right) \\
 &= \det \left( \begin{bmatrix} \lambda - 2 & 0 \\ -1 & \lambda - 1 \end{bmatrix} \right) \\
 &= (\lambda - 2) \cdot (\lambda - 1) - 0 \cdot (-1) \\
 &= (\lambda - 2) \cdot (\lambda - 1)
 \end{aligned}$$

The roots of the polynomial are

$$\lambda_1 = 2$$

$$\lambda_2 = 1$$

The eigenvectors associated to  $\lambda_1 = 2$  are the vectors

$$x_1 = \begin{bmatrix} x_{11} \\ x_{21} \end{bmatrix}$$

that solve the equation

$$\begin{bmatrix} \lambda_1 - 2 & 0 \\ -1 & \lambda_1 - 1 \end{bmatrix} \begin{bmatrix} x_{11} \\ x_{21} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

or

$$\begin{bmatrix} 0 & 0 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} x_{11} \\ x_{21} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

The last equation implies that

$$x_{11} = x_{21}$$

Therefore, the eigenspace of  $\lambda_1$  is the linear space that contains all vectors  $x_1$  of the form

$$x_1 = \alpha \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

where  $\alpha$  can be any scalar. Thus, the eigenspace of  $\lambda_1$  is generated by a single vector

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Therefore, it has dimension 1. As a consequence, the geometric multiplicity of  $\lambda_1$  is 1.

**Example** Consider the  $2 \times 2$  matrix

$$A = \begin{bmatrix} 1 & 0 \\ -1 & 1 \end{bmatrix}$$

The characteristic polynomial is

$$\begin{aligned} f(\lambda) &= \det \left( \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \begin{bmatrix} 1 & 0 \\ -1 & 1 \end{bmatrix} \right) \\ &= \det \left( \begin{bmatrix} \lambda - 1 & 0 \\ 1 & \lambda - 1 \end{bmatrix} \right) \\ &= (\lambda - 1) \cdot (\lambda - 1) - 0 \cdot 1 \\ &= (\lambda - 1) \cdot (\lambda - 1) \end{aligned}$$

and its roots are

$$\lambda_1 = 1$$

$$\lambda_2 = 1$$

Thus, there is a repeated eigenvalue ( $\lambda_1 = \lambda_2 = 1$ ) with algebraic multiplicity equal to 2. Its associated eigenvectors

$$x_1 = \begin{bmatrix} x_{11} \\ x_{21} \end{bmatrix}$$

solve the equation

$$\begin{bmatrix} \lambda_1 - 1 & 0 \\ 1 & \lambda_1 - 1 \end{bmatrix} \begin{bmatrix} x_{11} \\ x_{21} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

or

$$\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_{11} \\ x_{21} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

The equation is satisfied for  $x_{11} = 0$  and any value of  $x_{21}$ . As a consequence, the eigenspace of  $\lambda_1$  is the linear space that contains all vectors  $x_1$  of the form

$$x_1 = \alpha \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

where  $\alpha$  can be any scalar. Since the eigenspace of  $\lambda_1$  is generated by a single vector

$$\begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

it has dimension 1. As a consequence, the geometric multiplicity of  $\lambda_1$  is 1, less than its algebraic multiplicity, which is equal to 2.

**Example** Define the  $2 \times 2$  matrix

$$A = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

The characteristic polynomial is

$$\begin{aligned} f(\lambda) &= \det \left( \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \right) \\ &= \det \left( \begin{bmatrix} \lambda - 2 & 0 \\ 0 & \lambda - 2 \end{bmatrix} \right) \\ &= (\lambda - 2) \cdot (\lambda - 2) - 0 \cdot 0 \\ &= (\lambda - 2) \cdot (\lambda - 2) \end{aligned}$$

and its roots are

$$\lambda_1 = 2$$

$$\lambda_2 = 2$$

Thus, there is a repeated eigenvalue ( $\lambda_1 = \lambda_2 = 2$ ) with algebraic multiplicity equal to 2. Its associated eigenvectors

$$x_1 = \begin{bmatrix} x_{11} \\ x_{21} \end{bmatrix}$$

solve the equation

$$\begin{bmatrix} \lambda_1 - 2 & 0 \\ 0 & \lambda_1 - 2 \end{bmatrix} \begin{bmatrix} x_{11} \\ x_{21} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

or

$$\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_{11} \\ x_{21} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

The equation is satisfied for any value of  $x_{11}$  and  $x_{21}$ . As a consequence, the eigenspace of  $\lambda_1$  is the linear space that contains all vectors  $x_1$  of the form

$$x_1 = x_{11} \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} + x_{21} \cdot \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

where  $x_{11}$  and  $x_{21}$  are scalars that can be arbitrarily chosen. Thus, the eigenspace of  $\lambda_1$  is generated by the two linearly independent vectors

$$\begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Hence, it has dimension 2. As a consequence, the geometric multiplicity of  $\lambda_1$  is 2, equal to its algebraic multiplicity.

A takeaway message from the previous examples is that the algebraic and geometric multiplicity of an eigenvalue do not necessarily coincide.

## Relationship between algebraic and geometric multiplicity

The following proposition states an important property of multiplicities.

**Proposition** Let  $A$  be a  $K \times K$  matrix. Let  $\lambda_k$  be one of the eigenvalues of  $A$ . Then, the geometric multiplicity of  $\lambda_k$  is less than or equal to its algebraic multiplicity.

**Proof**

## Defective eigenvalues

When the geometric multiplicity of a repeated eigenvalue is strictly less than its algebraic multiplicity, then that eigenvalue is said to be **defective**.

An eigenvalue that is not repeated has an associated eigenvector which is different from zero. Therefore, the dimension of its eigenspace is equal to 1, its geometric multiplicity is equal to 1 and equals its algebraic multiplicity. Thus, an eigenvalue that is not repeated is also non-defective.

## Solved exercises

Below you can find some exercises with explained solutions.

### Exercise 1

Find whether the matrix

$$A = \begin{bmatrix} 1 & 2 \\ 0 & -1 \end{bmatrix}$$

has any defective eigenvalues.

**Solution**

### Exercise 2

Define



$$A = \begin{bmatrix} 1 & 0 & 2 \\ -1 & 1 & 3 \\ 0 & 0 & 2 \end{bmatrix}$$

Determine whether  $A$  possesses any defective eigenvalues.

**Solution**

## How to cite

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