

# Eigenvectors and Eigenvalues

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# Eigenvectors and Eigenvalues

Let  $V$  be a vector space and let

$$A : V \rightarrow V$$

be a linear map of  $V$  into itself. An element  $v \in V$  is called an eigenvector of  $A$  if there exists a number  $\lambda$  such that  $Av = \lambda v$ . If  $v \neq 0$  then  $\lambda$  is uniquely determined, because  $\lambda_1 v = \lambda_2 v$  implies  $\lambda_1 = \lambda_2$ . In this case, we say that  $\lambda$  is an eigenvalue of  $A$  belonging to the eigenvector  $v$ . We also say that  $v$  is an eigenvector with the eigenvalue  $\lambda$ . Instead of eigenvector and eigenvalue, one also uses the terms characteristic vector and characteristic value.

If  $A$  is a square  $n \times n$  matrix then an eigenvector of  $A$  is by definition an eigenvector of the linear map of  $\mathbb{R}^n$  into itself represented by this matrix. Thus an eigenvector  $X$  of  $A$  is a (column) vector of  $\mathbb{R}^n$  for which there exists  $\lambda \in \mathbb{R}$  such that  $AX = \lambda X$ .

Let  $V$  be the vector space over  $\mathbb{R}$  consisting of all infinitely differentiable functions. Let  $\lambda \in \mathbb{R}$ . Then the function  $f$  such that  $f(t) = e^{\lambda t}$  is an eigenvector of the derivative  $d/dt$  because  $df/dt = \lambda e^{\lambda t}$ .  
Let  $A : V \rightarrow V$  is a linear map, and  $v$  is an eigenvector of  $A$ , then for any non-zero scalar  $c$ ,  $cv$  is also an eigenvector of  $A$ , with the same eigenvalue.

## Theorem

*Let  $V$  be a vector space and let  $A : V \rightarrow V$  be a linear map. Let  $\lambda \in \mathbb{R}$ . Let  $V_\lambda$  be the subspace of  $V$  generated by all eigenvectors of  $A$  having  $\lambda$  as eigenvalue. Then every non-zero element of  $V_\lambda$  is an eigenvector of  $A$  having  $\lambda$  as an eigenvalue.*

## Proof.

Let  $v_1, v_2 \in V$  be such that  $Av_1 = \lambda v_1$  and  $Av_2 = \lambda v_2$ . Then

$$A(v_1 + v_2) = Av_1 + Av_2 = \lambda v_1 + \lambda v_2 = \lambda(v_1 + v_2).$$

If  $c \in K$  then  $A(cv_1) = cAv_1 = c\lambda v_1 = \lambda cv_1$ . This proves our theorem. □

The subspace  $V_\lambda$  is called the eigenspace of  $A$  belonging to  $\lambda$ .

## Theorem

*Let  $V$  be a vector space and let  $A : V \rightarrow V$  be a linear map. Let  $v_1, \dots, v_m$  be eigenvectors of  $A$ , with eigenvalues  $\lambda_1, \dots, \lambda_m$  respectively. Assume that these eigenvalues are distinct, i.e.*

$$\lambda_i \neq \lambda_j \quad \text{if} \quad i \neq j.$$

*Then  $v_1, \dots, v_m$  are linearly independent.*

Suppose  $V$  is a vector space of dimension  $n$  and  $A : V \rightarrow V$  is a linear map having  $n$  eigenvectors  $v_1, \dots, v_n$  whose eigenvalues  $\lambda_1, \dots, \lambda_n$  are distinct. Then  $\{v_1, \dots, v_n\}$  is a basis of  $V$ .

# The Characteristic Polynomial

## Theorem

*Let  $V$  be a finite dimensional vector space, and let  $\lambda$  be a number. Let  $A: V \rightarrow V$  be a linear map. Then  $\lambda$  is an eigenvalue of  $A$  if and only if  $A - \lambda I$  is not invertible.*

## Proof.

Assume that  $\lambda$  is an eigenvalue of  $A$ . Then there exists an element  $v \in V, v \neq 0$  such that  $Av = \lambda v$ . Hence  $Av - \lambda v = 0$ , and  $(A - \lambda I)v = 0$ . Hence  $A - \lambda I$  has a non-zero kernel, and  $A - \lambda I$  cannot be invertible. Conversely, assume that  $A - \lambda I$  is not invertible. We see that  $A - \lambda I$  must have a non-zero kernel, meaning that there exists an element  $v \in V, v \neq 0$  such that  $(A - \lambda I)v = 0$ . Hence  $Av - \lambda v = 0$ , and  $Av = \lambda v$ . Thus  $\lambda$  is an eigenvalue of  $A$ . This proves our theorem.  $\square$

Let  $A$  be an  $n \times n$  matrix,  $A = (a_{ij})$ . We define the characteristic polynomial  $P_A$  of  $A$  to be the determinant

$$P_A(t) = \text{Det}(tI - A).$$

We can also view  $A$  as a linear map from  $\mathbb{R}^n$  to  $\mathbb{R}^n$ , and we also say that  $P_A(t)$  is the characteristic polynomial of this linear map.



The characteristic polynomial of the matrix

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

is

$$\begin{vmatrix} t-1 & 1 & -3 \\ 2 & t-1 & -1 \\ 0 & -1 & t+1 \end{vmatrix}$$

which we expand according to the first column, to find

$$P_A(t) = t^3 - t^2 - 4t + 6.$$

For an arbitrary matrix  $A = (a_{ij})$ , the characteristic polynomial can be found by expanding according to the first column, and will always consist of a sum

$$(t - a_{11}) \cdots (t - a_{nn}) + \cdots .$$

Each term other than the one we have written down will have degree  $< n$ . Hence the characteristic polynomial is of type

$$P_A(t) = t^n + \text{terms of lower degree.}$$

## Theorem

*Let  $A$  be an  $n \times n$  matrix. A number  $\lambda$  is an eigenvalue of  $A$  if and only if  $\lambda$  is a root of the characteristic polynomial of  $A$ .*

This Theorem gives us an explicit way of determining the eigenvalues of a matrix, provided that we can determine explicitly the roots of its characteristic polynomial.

Find the eigenvalues and a basis for the eigenspaces of the matrix

$$\begin{pmatrix} 1 & 4 \\ 2 & 3 \end{pmatrix}.$$

Find the eigenvalues and a basis for the eigenspaces of the matrix

$$\begin{pmatrix} 2 & 1 & 0 \\ 0 & 1 & -1 \\ 0 & 2 & 4 \end{pmatrix}.$$

Find the eigenvalues and a basis for the eigenspaces of the matrix

$$\begin{pmatrix} 1 & 1 & -1 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix}.$$

## Theorem

*Let  $A, B$  be two  $n \times n$  matrices, and assume that  $B$  is invertible. Then the characteristic polynomial of  $A$  is equal to the characteristic polynomial of  $B^{-1}AB$ .*

# Eigenvalues and Eigenvectors of Symmetric Matrices

## Theorem

*Let  $A$  be a real symmetric matrix and let  $\lambda$  be an eigenvalue in  $\mathbb{C}$ . Then  $\lambda$  is real. If  $Z \neq O$  is a complex eigenvector with eigenvalue  $\lambda$ , and  $Z = X + iY$  where  $X, Y \in \mathbb{R}^n$ , then both  $X, Y$  are real eigenvectors of  $A$  with eigenvalue  $\lambda$ , and  $X$  or  $Y \neq O$ .*

## Proof

Let  $Z = (z_1, \dots, z_n)^t$  with complex coordinates  $z_i$ . Then

$$Z \cdot \bar{Z} = \bar{Z} \cdot Z = \bar{Z}^t Z = \bar{z}_1 z_1 + \dots + \bar{z}_n z_n = |z_1|^2 + \dots + |z_n|^2 > 0.$$

By hypothesis, we have  $AZ = \lambda Z$ . Then

$$\bar{Z}^t AZ = \bar{Z}^t \lambda Z = \lambda \bar{Z}^t Z.$$

The transpose of a  $1 \times 1$  matrix is equal to itself, so we also get

Proof.

$$Z^t A^t \bar{Z} = Z^t A \bar{Z} = \lambda \bar{Z}^t Z.$$

But  $\bar{A}Z = \bar{A}\bar{Z} = A\bar{Z}$  and  $\bar{A}Z = \bar{\lambda}Z = \bar{\lambda}\bar{Z}$ . Therefore

$\lambda \bar{Z}^t Z = \bar{\lambda} Z^t \bar{Z}$ . Since  $Z^t \bar{Z} \neq 0$  it follows that  $\lambda = \bar{\lambda}$ , so  $\lambda$  is real.

Now from  $AZ = \lambda Z$  we get

$$AX + iAY = \lambda X + i\lambda Y,$$

and since  $A, X, Y$  are real it follows that  $AX = \lambda X$  and  $AY = \lambda Y$ . This proves the theorem. □