

# Math 54: Linear Algebra and Differential Equations

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## Eigenvalues and Eigenvectors

Suppose that  $T : V \rightarrow V$  is an arbitrary linear operator (this is a linear operator since we are starting and ending in the same vector space). Given a basis  $\mathcal{B}$  for  $V$ , we get a matrix  $[T]_{\mathcal{B} \rightarrow \mathcal{B}}$  for the linear transformation  $T$  with respect to the basis  $\mathcal{B}$ . For each such ordered basis  $\mathcal{B}$ , we get out a different matrix  $[T]_{\mathcal{B} \rightarrow \mathcal{B}}$ .

We could imagine asking the following question: Can we find a basis  $\mathcal{B}$  for which the form of  $[T]_{\mathcal{B} \rightarrow \mathcal{B}}$  is as simple as possible - in particular, so that this matrix is diagonal? If this matrix is diagonal, then if the  $(i, i)$  entry is  $\lambda_i$ , this is saying that  $T(v_i) = \lambda v_i$  where  $v_i$  is the  $i$ th basis vector in  $\mathcal{B}$ . So if we want to try to find a basis in which a linear transformation is diagonal, we want to search for nonzero vectors where  $T$  applied to that vector gives a multiple of the original vector.

This motivates the following definition.

**Definition:** Let  $T : V \rightarrow V$  be a linear operator. A nonzero vector  $v$  is called an **eigenvector with eigenvalue**  $\lambda$  if  $T(v) = \lambda v$ .

As an example, note that  $u = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  is an eigenvector of the matrix transformation defined by the matrix  $A = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$  (sending  $v \in \mathbb{R}^2$  to  $Av \in \mathbb{R}^2$ ) since

$$Au = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \end{bmatrix} = 3u$$

so that  $u$  is an eigenvector of the matrix multiplication linear transformation defined by  $A$ , with eigenvalue 3.

As another example, given the identity transformation  $I : V \rightarrow V$  that sends every vector to itself, every nonzero element of  $V$  is an eigenvector of  $I$  with eigenvalue 1. Given the zero transformation  $Z : V \rightarrow V$  that sends every vector in  $V$  to the zero vector, every nonzero element of  $V$  is an eigenvector of  $A$  with eigenvalue 0.

Finally, let  $C^\infty(\mathbb{R})$  denote all smooth functions on the real line (functions that have infinitely many derivatives). Then,  $f(x) = e^{\lambda x}$  is an eigenvector with eigenvalue  $\lambda$ , since

$$\frac{d}{dx}(e^{\lambda x}) = \lambda e^{\lambda x}$$

We will now restrict to the specific case of linear operators which are matrix transformations, which recall are linear transformations from a vector space to itself. For a matrix transformation to map from  $\mathbb{R}^n$  to  $\mathbb{R}^n$ , we must restrict to square matrices. So the problem is, given a square matrix  $A$ , can we find all eigenvectors of  $A$ ?

If  $A$  has an eigenvector  $v$ , then note that  $Av = \lambda v$ . Rewriting this in a more convenient way, we have that

$$Av - \lambda v = (A - \lambda I)v = 0$$

So  $v$  is an eigenvector of  $A$  with eigenvalue  $\lambda$  if and only if  $v$  is in the nullspace of  $A - \lambda I$ . This gives us a nice way of finding all eigenvectors.

If we recall the invertible matrix theorem, this is saying that  $\lambda$  is an eigenvalue of  $A$  if and only if  $A - \lambda I$  has nontrivial (nonzero) nullspace if and only if  $\det(A - \lambda I) = 0$ . So to search for eigenvectors, we need to find all  $\lambda$  such that

$$\det(A - \lambda I) = 0$$

and for each such  $\lambda$ , we can find all eigenvectors with that eigenvalue by finding the nullspace of  $A - \lambda I$ . Note in particular that if we denote by  $U_\lambda$  the set of all eigenvectors of  $A$  with eigenvalue  $\lambda$ , with the zero vector added to this set, then  $U_\lambda$  is a subspace of  $\mathbb{R}^n$ .

It is a fact that eigenvectors for distinct eigenvalues are linearly independent (a fact which can be proved with the Vandermonde matrix). So if we want to find a basis  $\mathcal{B}$  for which a matrix transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^n$  given by an  $n$  by  $n$  matrix  $A$  is diagonal, we need to find  $n$  distinct linearly independent eigenvectors (so for each eigenspace, find a basis, and put all the bases together and see if you get  $n$  vectors). If there is a basis of  $n$  distinct linearly independent eigenvectors for an  $n$  by  $n$  matrix  $A$ , then we say that  $A$  and its associated matrix multiplication transformation is **diagonalizable**.

To find all eigenvalues  $\lambda$ , which are all  $\lambda$  such that  $\det(A - \lambda I) = 0$ , we need to solve the equation  $\det(A - xI) = 0$  for  $x$ . The polynomial

$$\text{char}_A(x) = \det(A - xI)$$

is an important polynomial associated to the square matrix  $A$ , called the **characteristic polynomial**.

The set of eigenvalues as we have discussed is the set of roots of the characteristic polynomial  $\text{char}_A(x)$ . However, roots of polynomials may have multiplicity. We define the **algebraic multiplicity of an eigenvalue**  $\lambda$  to be its multiplicity as a root of the characteristic polynomial. The **geometric multiplicity of an eigenvalue**  $\lambda$  is the dimension of  $U_\lambda$  (the nullspace of  $A - \lambda I$ ). It is a general fact that **geometric multiplicity of an eigenvalue is always less than the algebraic multiplicity of an eigenvalue**.

Another important fact is that the trace of a matrix is the sum of the roots of the characteristic polynomial (so it is the sum of the eigenvalues repeated with their algebraic multiplicity) and the determinant of a matrix is the product of the roots of the characteristic polynomial (so it is the product of the eigenvalues repeated with their algebraic multiplicity).

## Problem 1

Find all eigenvalues and eigenvectors of the following matrices.

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

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

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

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

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

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