

## Worksheet 23: Eigenvalues and Eigenvectors (§§7.1, 7.2)

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Let  $V$  be a vector space of dimension  $n$ , and let  $T : V \rightarrow V$  be a linear transformation.

**Definition:** The scalar  $\lambda \in \mathbb{R}$  is an *eigenvalue* of  $T$  if there is a non-zero vector  $v \in V$  such that

$$T(v) = \lambda v;$$

such a vector  $v$  is called an *eigenvector* of  $T$ , with corresponding eigenvalue  $\lambda$ .

Note that, by definition, eigenvectors are required to be nonzero!

**Problem 1.** Show that for any scalar  $\lambda$ , the set

$$E_\lambda = \{v \in V : T(v) = \lambda v\}$$

is a subspace of  $V$ , called the *eigenspace* corresponding to  $\lambda$ . Then complete the statement: “ $\lambda$  is an eigenvalue of  $T$  if and only if  $E_\lambda$  is ...”

**Solution:** Let  $T : V \rightarrow V$  be a linear transformation, and let  $\lambda \in \mathbb{R}$ . Since  $T(\vec{0}) = \vec{0} = \lambda\vec{0}$ , we have  $\vec{0} \in E_\lambda$ . If  $v \in E_\lambda$  and  $c \in \mathbb{R}$ , then  $T(cv) = cT(v) = \lambda(cv)$ , so  $cv \in E_\lambda$ . Finally, if  $v, w \in E_\lambda$  then  $T(v + w) = T(v) + T(w) = \lambda v + \lambda w = \lambda(v + w)$ , so  $v + w \in E_\lambda$ . This shows that  $E_\lambda$  is a subspace of  $V$ . Furthermore,  $\lambda$  is an eigenvalue of  $T$  if and only if  $E_\lambda \neq \{\vec{0}\}$ .

**Problem 2.** For the following transformations, try to find an eigenvector using any methods you can think of, including basic geometry, if this is possible. What are the corresponding eigenvalues?

- (a)  $V = \mathbb{R}^2$ ,  $T$  = reflection over the  $x$ -axis.
- (b)  $V = \mathbb{R}^2$ ,  $T$  = reflection over the line  $x = y$ .
- (c)  $V = \mathbb{R}^2$ ,  $T$  = rotation by  $90^\circ$ .
- (d)  $V = \mathbb{R}^2$ ,  $T$  = left multiplication by  $\begin{bmatrix} 2 & 1 \\ 0 & 3 \end{bmatrix}$ .
- (e)  $V = \mathbb{R}^2$ ,  $T$  = left multiplication by  $\begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix}$ .
- (f)  $V = \mathcal{P}_3$  the space of polynomials of degree less than or equal 3 in the variable  $t$ ,

$$T(f) = f'.$$

- (g)  $V = \mathbb{R}^{2 \times 2}$ ,  $T$  is the zero transformation.

**Solution:**

- (a) eigenvectors  $\vec{e}_1$  and  $\vec{e}_2$ , with corresponding eigenvalues 1 and  $-1$
- (b) eigenvectors  $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$  and  $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$ , with corresponding eigenvalues 1 and  $-1$
- (c) no eigenvectors or eigenvalues
- (d) eigenvectors  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and  $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ , with corresponding eigenvalues 2 and 3
- (e) eigenvector  $\vec{e}_1$ , with corresponding eigenvalue 2
- (f) any constant function is an eigenvector, with eigenvalue 0
- (g) Any zero transformation has  $\lambda = 0$  as its only eigenvalue, with the entire space  $\mathbb{R}^{2 \times 2}$  the corresponding eigenspace.

**Problem 3.** Suppose that  $\lambda$  is an eigenvalue of the linear transformation  $T : V \rightarrow V$ .

- (a) Given  $n \in \mathbb{N}$ , is  $\lambda$  an eigenvalue of  $T^n$ ? Can you find any eigenvalues of  $T^n$ ?
- (b) Supposing that  $T$  is invertible, is  $\lambda$  an eigenvalue of  $T^{-1}$ ? Can you find any eigenvalues of  $T^{-1}$ ?

**Solution:**

- (a) If  $v$  is an eigenvector of  $T$  corresponding to  $\lambda$ , then  $T^n(v) = \lambda^n v$ , so  $v$  is an eigenvector of  $T^n$  with corresponding eigenvalue  $\lambda^n$ . Thus  $\lambda$  is not an eigenvalue of  $T^n$  unless there is some eigenvalue  $\mu$  of  $T$  such that  $\mu^n = \lambda$ .
- (b) If  $T$  is invertible and  $v$  is an eigenvector of  $T$  with corresponding eigenvalue  $\lambda$ , then  $T^{-1}(\lambda v) = v$ , so  $T^{-1}(v) = \frac{1}{\lambda}v$ , which shows that  $v$  is an eigenvector of  $T^{-1}$  with corresponding eigenvalue  $\lambda^{-1}$ . Thus  $\lambda$  is not an eigenvalue of  $T^{-1}$  unless  $\lambda^{-1}$  also happens to be an eigenvalue of  $T$ .

**Problem 4.** The *characteristic polynomial* of the linear transformation  $T : V \rightarrow V$  is the polynomial  $f_T$  (which we write here in the variable  $\lambda$ ) given by

$$f_T(\lambda) = \det(\lambda I - T),$$

where  $I : V \rightarrow V$  is the identity transformation. Here we are thinking of  $\lambda I - T$  as a new linear transformation from  $V$  to  $V$ , defined by

$$(\lambda I - T)(v) = \lambda v - T(v) \quad \text{for all } v \in V.$$

Recall that in order to compute the determinant of a linear transformation  $S : V \rightarrow V$ , we just compute the determinant of the  $\mathcal{B}$ -matrix  $[S]_{\mathcal{B}}$  of  $S$  where  $\mathcal{B}$  is *any* basis of  $V$ .

Write down the characteristic polynomials for all the linear transformations in Problem 2.

**Solution:**

(a)  $(\lambda - 1)(\lambda + 1)$

(b)  $(\lambda - 1)(\lambda + 1)$

(c)  $\lambda^2 + 1$

(d)  $(\lambda - 2)(\lambda - 3)$

(e)  $(\lambda - 2)^2$

(f)  $\lambda^4$

(g)  $\lambda^4$

**Problem 5: A systematic way of finding eigenvalues.** If  $v$  is an eigenvector of  $T$  with corresponding eigenvalue  $\lambda$ , then

$$(\lambda I - T)(v) = \vec{0},$$

so the linear transformation  $\lambda I - T$  has a nontrivial kernel, hence is not invertible, which means

$$\det(\lambda I - T) = 0.$$

Conversely, reversing the argument shows that if  $\det(\lambda I - T) = 0$ , then  $\lambda$  is an eigenvalue of  $T$ . This shows that *the eigenvalues of  $T$  are just the roots of the characteristic polynomial of  $T$ .*

Using this method, list *all* eigenvalues of the transformations in Problem 2, and thus verify that you didn't miss any the first time around. (Keep these eigenvalues handy for Problem 6 below).

**Solution:**

(a)  $\lambda = \pm 1$

(b)  $\lambda = \pm 1$

(c) none

(d)  $\lambda = 2, 3$

(e)  $\lambda = 2$

(f)  $\lambda = 0$

(g)  $\lambda = 0$

**Definition:** Let  $\lambda$  be an eigenvalue of  $T$ . The *algebraic multiplicity* “ $\text{almu}(\lambda)$ ” of  $\lambda$  is the number of times that  $\lambda$  occurs as a root of the characteristic polynomial  $f_T$  of  $T$ ; that is, if we write  $f_T$  in

the variable  $x$ , then  $\text{almu}(\lambda)$  is the largest power of  $(x - \lambda)$  that is a factor of  $f_T(x)$ . The *geometric multiplicity* “ $\text{gemu}(\lambda)$ ” of  $\lambda$  is the dimension of the corresponding eigenspace  $E_\lambda$ .

**Problem 6.** Compute the algebraic and geometric multiplicities of all the eigenvalues of the transformations in Problem 2. Do you find experimentally any relation between the algebraic and geometric multiplicities?

**Solution:** Every eigenvalue in Problem 2 has geometric multiplicity 1, except the eigenvalue  $\lambda = 0$  in part (g), which has geometric multiplicity 4. The eigenvalue  $\lambda = 2$  in part (e) has algebraic multiplicity 2, the eigenvalue  $\lambda = 0$  in parts (f) and (g) have algebraic multiplicity 4, and all the other eigenvalues in Problem 2 have algebraic multiplicity 1. For each eigenvalue, the geometric multiplicity is always less than or equal to the algebraic multiplicity.

**Problem 7.** Suppose  $v_1$  and  $v_2$  are eigenvectors of  $T$  that correspond to *distinct* eigenvalues  $\lambda_1$  and  $\lambda_2$ , respectively. Show that  $v_1$  and  $v_2$  are linearly independent. What is  $E_{\lambda_1} \cap E_{\lambda_2}$ ?

**Solution:** Let  $\vec{v}_1$  and  $\vec{v}_2$  be eigenvectors of  $T$  corresponding to distinct eigenvalues  $\lambda_1$  and  $\lambda_2$ , and suppose  $c_1\vec{v}_1 + c_2\vec{v}_2 = \vec{0}$ . Applying  $T$  to this equation gives

$$c_1\lambda_1\vec{v}_1 + c_2\lambda_2\vec{v}_2 = \vec{0},$$

and scaling by  $\lambda_1$  gives

$$c_1\lambda_1\vec{v}_1 + c_2\lambda_1\vec{v}_2 = \vec{0}.$$

Subtracting the second of these equations from the first, we obtain

$$c_2(\lambda_2 - \lambda_1)\vec{v}_2 = \vec{0}.$$

Since  $\lambda_1 \neq \lambda_2$  and  $\vec{v}_2 \neq \vec{0}$ , we see that  $c_2 = 0$ , and a similar argument shows  $c_1 = 0$ . Therefore  $(\vec{v}_1, \vec{v}_2)$  is linearly independent.

Furthermore, it follows immediately from the definition of eigenspace that  $E_{\lambda_1} \cap E_{\lambda_2} = \{\vec{0}\}$  whenever  $\lambda_1 \neq \lambda_2$ , since if  $\vec{v} \in E_{\lambda_1} \cap E_{\lambda_2}$  where  $\lambda_1 \neq \lambda_2$ , then  $\lambda_1\vec{v} = T(\vec{v}) = \lambda_2\vec{v}$ , which implies  $\vec{v} = \vec{0}$ . (This gives another way of showing that  $\vec{v}_1$  and  $\vec{v}_2$  are linearly independent).

**Problem 8.** Extend the statement and proof of the previous problem to  $r$  eigenvectors.

**Solution:** We show that if  $\vec{v}_1, \dots, \vec{v}_r$  are eigenvectors of  $T$  corresponding to distinct eigenvalues  $\lambda_1, \dots, \lambda_r$ , respectively, then  $(\vec{v}_1, \dots, \vec{v}_r)$  is linearly independent.

We argue by induction on  $r$ . The base case  $r = 1$  follows immediately from the fact that eigenvectors are nonzero. For the inductive step, let  $r > 1$ , suppose the claim holds for any set of  $r - 1$  eigenvectors with pairwise distinct eigenvalues, and let  $\vec{v}_1, \dots, \vec{v}_r$  be eigenvectors of  $T$  with corresponding pairwise distinct eigenvalues  $\lambda_1, \dots, \lambda_r$ , respectively. Suppose  $\sum_{i=1}^r c_i\vec{v}_i = \vec{0}$ . Separately applying  $T$  to this equation and multiplying it by  $\lambda_r$ , as in the proof of (6), we obtain

$$\sum_{i=1}^r c_i\lambda_i\vec{v}_i = \vec{0} = \sum_{i=1}^r c_i\lambda_r\vec{v}_i,$$

so we have  $\sum_{i=1}^{r-1} c_i(\lambda_i - \lambda_r)\vec{v}_i = \vec{0}$  after subtracting. Using the fact that  $\lambda_i \neq \lambda_r$  for each  $i \neq r$ , and applying the inductive hypothesis, we see that  $c_i = 0$  for each  $1 \leq i < r$ . Therefore  $c_r\vec{v}_r = \vec{0}$  (from the original equation), which implies  $c_r = 0$  since  $\vec{v}_r \neq \vec{0}$ . Thus each  $c_i$  is zero, showing that  $(\vec{v}_1, \dots, \vec{v}_r)$  is linearly independent as claimed, completing the induction.

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*Everything on this worksheet has been defined in terms of a linear transformation  $T : V \rightarrow V$ . If  $A$  is an  $n \times n$  matrix, then the map  $T_A(\vec{x}) = A\vec{x}$  is a linear transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^n$ . Thus we may redefine all the notions introduced in this worksheet in the context of matrices: the eigenvalues, eigenvectors, and characteristic polynomial of a square matrix  $A$  are just the eigenvalues, eigenvectors, and characteristic polynomial of the transformation  $T_A$ .*