

5.1 Eigenvectors and Eigenvalues

For this section we will assume that A is an $n \times n$ matrix. So any transformation $T(x) = Ax$ sends vectors from \mathbb{R}^n to \mathbb{R}^n .

Example 5.1.1. Consider the linear transformation $T(x) = Ax$ defined by $A = \begin{bmatrix} 4 & 1 \\ -2 & 1 \end{bmatrix}$.

1. Describe what this transformation does to the standard basis vectors in \mathbb{R}^2

$$A \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 4 & 1 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 4 \\ -2 \end{bmatrix} \quad A \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 4 & 1 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

2. Let $b_1 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$. Calculate $T(b_1)$ and describe what happens.

$$\begin{bmatrix} 4 & 1 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} -4 + 1 \\ 2 + 1 \end{bmatrix} = \begin{bmatrix} -3 \\ 3 \end{bmatrix} = 3 \begin{bmatrix} -1 \\ 1 \end{bmatrix} = 3 b_1$$

eigenvector

eigenvalue

3. Let $b_2 = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$. Calculate $T(b_2)$ and describe what happens.

$$\begin{bmatrix} 4 & 1 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \begin{bmatrix} -4 + 2 \\ 2 + 2 \end{bmatrix} = \begin{bmatrix} -2 \\ 4 \end{bmatrix} = 2 \begin{bmatrix} -1 \\ 2 \end{bmatrix} = 2 b_2$$

$$A \begin{bmatrix} -2 \\ 4 \end{bmatrix} = \begin{bmatrix} -4 \\ 8 \end{bmatrix}$$

eigenvalue

eigenvector

Eigenvectors and Eigenvalues

Definition 5.1. An **eigenvector** of an $n \times n$ matrix A is a nonzero vector x such that $Ax = \lambda x$ for some scalar λ . The scalar λ is called an **eigenvalue** of A if there is a nontrivial solution to the equation $Ax = \lambda x$. (Note that an eigenvector must be nonzero but eigenvalues can be zero.)

λ is unique
 x is not

In this section we will be given either an eigenvalue or an eigenvector for each problem.

$Ax - \lambda x = 0$ can't be done this way

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

show this

Example 5.1.2. Let $A = \begin{bmatrix} 3 & 2 \\ 3 & 8 \end{bmatrix}$. Show that $b = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$ is an eigenvector of A . Determine the corresponding eigenvalue.

Eigen vector

$A b = \lambda b$
 λ is the eigenvalue

$$\begin{bmatrix} 3 & 2 \\ 3 & 8 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \end{bmatrix} = \begin{bmatrix} 9 \\ 27 \end{bmatrix} = 9 \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

$$A b = 9 b$$

$\lambda = 9$ is the eigenvalue

Solve for b

Example 5.1.3. Let $A = \begin{bmatrix} 3 & 2 \\ 3 & 8 \end{bmatrix}$. Show that $\lambda = 2$ is an eigenvalue of A . Determine eigenvector whose eigenvalue is 2.

$$b = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

$A b = 2 b$

$$\begin{bmatrix} 3 & 2 \\ 3 & 8 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = 2 \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

Solve

$$\begin{cases} 3b_1 + 2b_2 = 2b_1 \\ 3b_1 + 8b_2 = 2b_2 \end{cases}$$

$$\begin{cases} b_1 + 2b_2 = 0 \\ 3b_1 + 6b_2 = 0 \end{cases}$$

$$\rightarrow \begin{bmatrix} 1 & 2 & 0 \\ 3 & 6 & 0 \end{bmatrix} \xrightarrow{RR}$$

$$\xrightarrow{RR} \begin{bmatrix} 1 & 2 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{matrix} b_1 + 2b_2 = 0 \\ b_1 = -2b_2 \\ b_2 = b_2 \end{matrix}$$

$$\begin{bmatrix} 3b_1 + 2b_2 \\ 3b_1 + 8b_2 \end{bmatrix} = \begin{bmatrix} 2b_1 \\ 2b_2 \end{bmatrix}$$

$A b - 2I b = 0$

$$\begin{bmatrix} 3 & 2 \\ 3 & 8 \end{bmatrix} b - 2 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} b = 0$$

$$\begin{bmatrix} 3-2 & 2 \\ 3 & 8-2 \end{bmatrix} b = 0$$

$b = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} -2 \\ 1 \end{bmatrix} b_2$

Eigenspace
Definition 5.2. The set of all solutions to $(A - \lambda I)x = 0$ is a subspace of \mathbb{R}^n . It is called the **eigenspace** of A corresponding to the eigenvalue λ .

Example 5.1.4. Let $A = \begin{bmatrix} 6 & 3 & -4 \\ 2 & 7 & -4 \\ 2 & 3 & 0 \end{bmatrix}$. if $\lambda = 4$ is an eigenvalue of A find a basis for the eigenspace of A .

Solve $(A - \lambda I)x = 0$

$$\left(\begin{bmatrix} 6 & 3 & -4 \\ 2 & 7 & -4 \\ 2 & 3 & 0 \end{bmatrix} - 4 \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \right) \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = 0$$

$$\begin{bmatrix} 2 & 3 & -4 \\ 2 & 3 & -4 \\ 2 & 3 & -4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} 2 & 3 & -4 & 0 \\ 2 & 3 & -4 & 0 \\ 2 & 3 & -4 & 0 \end{bmatrix} \xrightarrow{RR} \begin{bmatrix} 1 & 3/2 & -2 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$x_1 + \frac{3}{2}x_2 - 2x_3 = 0$$

$$x_1 = -\frac{3}{2}x_2 + 2x_3$$

$$x_2 = x_2$$

$$x_3 = x_3$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -3/2 \\ 1 \\ 0 \end{bmatrix} x_2 + \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix} x_3$$

Basis = $\left\{ \begin{bmatrix} -3/2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix} \right\}$
for $\lambda = 4$

Eigenvector Linear Independence Theorem

Theorem 5.1. If v_1, \dots, v_r are eigenvectors that correspond to distinct eigenvalues $\lambda_1, \dots, \lambda_r$ of an $n \times n$ matrix A , then the set $\{v_1, \dots, v_r\}$ is linearly independent.

$$x_1 = x_1$$

$$x_2 = x_2$$

$$x_3 = \frac{1}{2}x_1 + \frac{3}{4}x_2$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 1/2 \end{bmatrix} x_1 + \begin{bmatrix} 0 \\ 1 \\ 3/4 \end{bmatrix} x_2$$

$E_{\lambda=4}$

Example 5.1.5. A is an $m \times n$ matrix. Mark each statement TRUE or FALSE. $x \neq 0$

1. If $Ax = \lambda x$ for some vector x , then λ is an eigenvalue of A .

True

2. A matrix A is not invertible if and only if 0 is an eigenvalue of A ($Ax = 0x$).

True

3. A number c is an eigenvalue of A if and only if the equation $(A - cI)x = 0$ has a nontrivial solution.

$x=0$ is NEVER an eigenvector

4. To find the eigenvalues of A , reduce A to echelon form.

No solve $(A - \lambda I)x = 0$

5. If $Ax = \lambda x$ for some scalar λ , then x is an eigenvector of A .

True

6. If v_1 and v_2 are linearly independent eigenvectors, then they correspond to distinct eigenvalues.

False see ex 5.1.4

7. An eigenspace of A is a null space of a certain matrix.

True (Matrix) $x=0$ is null space

8. An $n \times n$ matrix can have at most n eigenvalues.

True

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