

Section 5.1

3.
$$\begin{bmatrix} -3 & 1 \\ -3 & 8 \end{bmatrix} \begin{bmatrix} 1 \\ 4 \end{bmatrix} = \begin{bmatrix} 1 \\ 29 \end{bmatrix}$$
 since $\begin{bmatrix} 1 \\ 29 \end{bmatrix}$ is not a scalar multiple of $\begin{bmatrix} 1 \\ 4 \end{bmatrix}$, $\begin{bmatrix} 1 \\ 4 \end{bmatrix}$ is not an eigenvector.

6.
$$\begin{bmatrix} 3 & 6 & 7 \\ 3 & 3 & 7 \\ 5 & 6 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} = \begin{bmatrix} -2 \\ 4 \\ -2 \end{bmatrix} = -2 \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$$

$\begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$ is an eigenvector for the matrix corresponding to the eigenvalue -2 .

9.
$$A = \begin{bmatrix} 5 & 0 \\ 2 & 1 \end{bmatrix}, \lambda = 1, 5.$$

To find a basis for the eigenspace corresponding to $\lambda = 1$, we find a basis for the null space of $A - I$.

$$A - I = \begin{bmatrix} 5 & 0 \\ 2 & 1 \end{bmatrix} - \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 4 & 0 \\ 2 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

Thus $x_1 = 0$, and x_2 is free.

$\vec{x} = x_2 \begin{bmatrix} 0 \\ 1 \end{bmatrix}$, $x_2 \in \mathbb{R}$, are the solutions

to $(A - I)\vec{x} = \vec{0}$.

Since $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$ is a basis for the nullspace of $A - I$, $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$ is a basis for the eigenspace of A corresponding to $\lambda = 1$.

To find a basis for the eigenspace of A corresponding to $\lambda = 5$, we find a basis for the nullspace of $A - 5I$.

$$A - 5I = \begin{bmatrix} 5 & 0 \\ 2 & 1 \end{bmatrix} - \begin{bmatrix} 5 & 0 \\ 0 & 5 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 2 & -4 \end{bmatrix} \sim \begin{bmatrix} 1 & -2 \\ 0 & 0 \end{bmatrix}$$

$$\begin{array}{l} x_1 = 2x_2 \\ x_2 = x_2 \end{array} \quad \vec{x} = x_2 \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \quad x_2 \in \mathbb{R}.$$

Thus $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$ is a basis for the eigenspace of A corresponding to $\lambda = 5$.

$$\begin{aligned} \underline{14.} \quad A - (-2I) &= \begin{bmatrix} 3 & 0 & -1 \\ 1 & -1 & 0 \\ 4 & -13 & 3 \end{bmatrix} \sim \begin{bmatrix} 1 & -1 & 0 \\ 3 & 0 & -1 \\ 4 & -13 & 3 \end{bmatrix} \\ &\sim \begin{bmatrix} 1 & -1 & 0 \\ 0 & 3 & -1 \\ 0 & -9 & 3 \end{bmatrix} \sim \begin{bmatrix} 1 & -1 & 0 \\ 0 & 3 & -1 \\ 0 & 0 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & -\frac{1}{3} \\ 0 & 1 & -\frac{1}{3} \\ 0 & 0 & 0 \end{bmatrix} \end{aligned}$$

$$\begin{aligned} x_1 &= \frac{1}{3}x_3 \\ x_2 &= \frac{1}{3}x_3 \\ x_3 &= x_3 \end{aligned} \quad \vec{x} = x_3 \begin{bmatrix} \frac{1}{3} \\ \frac{1}{3} \\ 1 \end{bmatrix}, \quad x_3 \in \mathbb{R}.$$

So $\begin{bmatrix} 1 \\ 1 \\ 3 \end{bmatrix}$ is a basis for the eigenspace of A corresponding to $\lambda = -2$.

17. Since the matrix is upper triangular, Theorem 1 states the eigenvalues are the diagonal entries. Thus the eigenvalues are $0, 2, -1$.

19.

$$\begin{bmatrix} 1 & 2 & 3 \\ 1 & 2 & 3 \\ 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 6 \\ 6 \\ 6 \end{bmatrix} = 6 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

The sum of the rows are all 6, so 6 is an eigenvalue.

The columns are all multiples of $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ so the matrix is not invertible. Thus 0 is an eigenvalue.

(You just need to identify one of the eigenvalues.)

23.

An $n \times n$ matrix can have at most n distinct eigenvalues because of Theorem 2. The eigenvectors corresponding to distinct eigenvalues are linearly independent. Since the dimension of \mathbb{R}^n is n , we can have at most n vectors, in \mathbb{R}^n , in a linearly independent set. Thus an $n \times n$ matrix can have at most n distinct eigenvalues.

25.

Since λ is an eigenvalue of A , there exists an eigenvector \vec{x} satisfying, $A\vec{x} = \lambda\vec{x}$.

Since A is invertible, we may multiply A^{-1} on the left of each side of the equation.

$$A^{-1}(A\vec{x}) = A^{-1}(\lambda\vec{x})$$

$$I\vec{x} = A^{-1}(\lambda\vec{x})$$

$$\vec{x} = A^{-1}\lambda\vec{x}$$

Since A is invertible, $\lambda \neq 0$.

$$\frac{1}{\lambda}\vec{x} = \frac{1}{\lambda}A^{-1}\lambda\vec{x} = A^{-1}\vec{x},$$

$$\text{Thus } A^{-1}\vec{x} = \lambda^{-1}\vec{x}.$$

Since \vec{x} is an eigenvector, $\vec{x} \neq \vec{0}$. Thus

λ^{-1} is an eigenvalue of A^{-1} .

26. Let λ be an eigenvalue of A , and suppose $A^2 = 0$.

Since λ is an eigenvalue of A , there exists a nonzero vector \vec{x} such that

$$A\vec{x} = \lambda\vec{x}.$$

Multiply A on the left of each side of the equation.

$$A(A\vec{x}) = A(\lambda \vec{x})$$

On one hand,

$$A(A\vec{x}) = A^2 \vec{x} = 0 \vec{x} = \vec{0}.$$

On the other hand,

$$A(\lambda \vec{x}) = \lambda(A\vec{x}) = \lambda(\lambda \vec{x}) = \lambda^2 \vec{x}.$$

Thus $\lambda^2 \vec{x} = \vec{0}$. Since \vec{x} is an eigenvector, $\vec{x} \neq \vec{0}$. Thus $\lambda^2 = 0$, and $\lambda = 0$.

Since λ was an arbitrary eigenvalue of A , all eigenvalues of A are equal to 0.

27. Recall that λ is an eigenvalue of A if and only if $A - \lambda I$ has a non-trivial nullspace. $A - \lambda I$ has a non-trivial nullspace if and only if $A - \lambda I$ is not invertible. (Invertible Matrix Theorem). $A - \lambda I$ is not invertible if and only if $(A - \lambda I)^T$ is not invertible. Note $(A - \lambda I)^T = A^T - (\lambda I)^T = A^T - \lambda I$.

Thus $A - \lambda I$ is not invertible if and only if $A^T - \lambda I$ is not invertible. $A^T - \lambda I$ is not invertible if and only if $A^T - \lambda I$ has a non-trivial null space. $A^T - \lambda I$ has a nontrivial null space if and only if λ is an eigenvalue of A^T .

29. Let A be a matrix whose row sums are all equal to s . Let \vec{x} be the all-ones vector. Then $A\vec{x} = s\vec{x}$. Thus s is an eigenvalue of A .