

Summary: Solving Systems of Equations

$$\begin{array}{cccccc} a_{11}x_1 & + & a_{12}x_2 & + & \cdots & + & a_{1n}x_n & = & b_1 \\ a_{21}x_1 & + & a_{22}x_2 & + & \cdots & + & a_{2n}x_n & = & b_2 \\ \vdots & & \vdots & & \vdots & & \vdots & & \vdots \\ a_{m1}x_1 & + & a_{m2}x_2 & + & \cdots & + & a_{mn}x_n & = & b_m \end{array}$$

$$A\mathbf{x} = \mathbf{b}$$

1. If A is not square (i.e., more unknowns than equations or more equations than unknowns), then augmented matrix \rightarrow row echelon or reduced row echelon form.

2. If A is square, then

$(A|b) \rightarrow$ row echelon form & back substitute

$(A|b) \rightarrow$ reduced row echelon form

find A^{-1} , if possible, then $\mathbf{x} = A^{-1}\mathbf{b}$

Cramer's rule: $x_i = \frac{\det A_i}{\det A}$, provided

$\det A \neq 0$

Note: If A^{-1} does not exist, or

if $\det A = 0$, then the system either

has infinitely many solutions or no solutions.

BUT, Special Case:

If the system is homogeneous, then it has infinitely many solutions; "no solutions" is not an option for homogeneous systems,

Section 5.7. Vector Spaces

$\mathbb{R}^2 = \{(a, b) : a, b \in \mathbb{R}\}$ – "the plane"

$\mathbb{R}^3 = \{(a, b, c) : a, b, c \in \mathbb{R}\}$ – "3-space"

$$\mathbb{R}^4 = \{(x_1, x_2, x_3, x_4)\} - \text{“4-space”}$$

$$\mathbb{R}^n = \{(x_1, x_2, x_3, \dots, x_n)\} \quad \text{ordered}$$

n -tuples of real numbers

For any two vectors $\mathbf{u} = (a_1, a_2, \dots, a_n)$
and $\mathbf{v} = (b_1, b_2, \dots, b_n)$ in \mathbb{R}^n , we
have

$$\begin{aligned}\mathbf{u} + \mathbf{v} &= (a_1, a_2, \dots, a_n) + (b_1, b_2, \dots, b_n) \\ &= (a_1 + b_1, a_2 + b_2, \dots, a_n + b_n)\end{aligned}$$

and for any real number λ ,

$$\begin{aligned}\lambda \mathbf{v} &= \lambda (a_1, a_2, \dots, a_n) \\ &= (\lambda a_1, \lambda a_2, \dots, \lambda a_n).\end{aligned}$$

Clearly, the sum of two vectors in
 \mathbb{R}^n is another vector in \mathbb{R}^n and a

scalar multiple of a vector in \mathbb{R}^n is
a vector in \mathbb{R}^n .

Properties:

Let u, v and w be vectors in \mathbb{R}^n .

Addition:

1. $u + v \in \mathbb{R}^n$ **closed**

2. $u + v = v + u$ **commutative**

3. $u + (v + w) = (u + v) + w$

associative

4. **Zero vector:** $\mathbf{0} = (0, 0, 0, \dots, 0)$,

$\mathbf{v} + \mathbf{0} = \mathbf{0} + \mathbf{v} = \mathbf{v}$ (additive identity).

5. **Additive Inverse:** For each vector $\mathbf{v} \in \mathbb{R}^n$, there is a unique vector $-\mathbf{v}$ such that

$$\mathbf{v} + (-\mathbf{v}) = \mathbf{v} - \mathbf{v} = \mathbf{0}$$

$-\mathbf{v}$ is the additive inverse (or negative) of \mathbf{v} .

Multiplication by a scalar:

If α and β are numbers, and \mathbf{u} and \mathbf{v} are vectors, then:

1. $\alpha \mathbf{v} \in \mathbb{R}^n$ (closed)
2. $1\mathbf{v} = \mathbf{v}$ (1 multiplicative identity)
3. $\alpha(\beta \mathbf{v}) = (\alpha\beta) \mathbf{v}$ (associative property)
4. $(\alpha + \beta)\mathbf{v} = \alpha \mathbf{v} + \beta \mathbf{v}$ (distributive property)

5. $\alpha(\mathbf{u} + \mathbf{v}) = \alpha\mathbf{u} + \alpha\mathbf{v}$ (distributive property)

Any non-empty set V on which there are defined two operations, addition ($+$) and multiplication by a scalar, which satisfy the properties 1 - 5 for addition and 1 - 5 for multiplication by a scalar is called a **vector space**.

Examples:

1. \mathbb{R}^n

2. $\mathcal{C}(0, 1) = \{f : (0, 1) \rightarrow \mathbb{R} \mid f \text{ is continuous}\}$

3. The set \mathcal{S} of solutions of

$$y'' + p(x)y' + q(x)y = 0$$

Section 5.8. Linear Dependence/Independence in \mathbb{R}^n

Let V be a vector space and let

$$\{v_1, v_2, v_3, \dots, v_k\}$$

be a set of vectors in V . Let

$$\{c_1, c_2, c_3, \dots, c_k\}$$

be real numbers. Then

$$v = c_1v_1 + c_2v_2 + c_3v_3 + \dots + c_kv_k$$

is a **linear combination** of v_1, \dots, v_k .

$\mathbb{R}^2 = \{(a, b) : a, b \in \mathbb{R}\}$ – "the plane"

$\mathbb{R}^3 = \{(a, b, c) : a, b, c \in \mathbb{R}\}$ – "3-space"

Let $\mathcal{S} = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ be a set of vectors in \mathbb{R}^n .

The set \mathcal{S} is **linearly dependent** if there exist k numbers c_1, c_2, \dots, c_k **NOT ALL ZERO** such that

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_k\mathbf{v}_k = \mathbf{0}.$$

$(c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_k\mathbf{v}_k$ is a **linear combination** of $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k)$

\mathcal{S} is **linearly independent** if it is not linearly dependent. That is, \mathcal{S} is linearly independent if

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \cdots + c_k\mathbf{v}_k = \mathbf{0}$$

implies $c_1 = c_2 = \cdots = c_k = 0$.

The set \mathcal{S} is **linearly dependent** if there exist k numbers c_1, c_2, \dots, c_k NOT ALL ZERO such that

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_k\mathbf{v}_k = \mathbf{0}.$$

Another way to say this:

The set \mathcal{S} is **linearly dependent** if one of the vectors can be written as a linear combination of the other vectors.

The set \mathcal{S} is **linearly dependent** if there exist k numbers c_1, c_2, \dots, c_k NOT ALL ZERO such that

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_k\mathbf{v}_k = \mathbf{0}.$$

NOTE: If there is one such set

$$\{c_1, c_2, c_3, \dots, c_k\},$$

then there are **infinitely many** such sets.

1. Two vectors v_1, v_2 .

Linearly dependent iff one vector is a constant multiple of the other.

Examples:

$$v_1 = (1, -2, 4), \quad v_2 = \left(-\frac{1}{2}, 1, -2\right)$$

linearly dependent: $v_1 = -2v_2$

$$\mathbf{v}_1 = (2, -4, 5), \quad \mathbf{v}_2 = (0, 0, 0)$$

linearly dependent: $\mathbf{v}_2 = 0\mathbf{v}_1$

$$\mathbf{v}_1 = (5, -2, 0), \quad \mathbf{v}_2 = (-3, 1, 9)$$

linearly independent

2. Given the three vectors v_1, v_2, v_3 in \mathbb{R}^2 :

$$v_1 = (1, -1), \quad v_2 = (-2, 3)$$

$$v_3 = (3, -5)$$

1. $\{v_1, v_2\}$ Dependent or independent??

Solution 1. Independent: v_1 is not a constant multiple of v_2 ; and v_2 is not a constant multiple of v_1 .

Solution 2. Suppose they are dependent. Then there are two numbers c_1 , c_2 , not both 0, such that

$$c_1(1, -1) + c_2(-2, 3) = \mathbf{0}.$$

That is

$$c_1 - 2c_2 = 0$$

$$-c_1 + 3c_2 = 0$$

$$\mathbf{v}_1 = (1, -1), \mathbf{v}_2 = (-2, 3)$$

$$\mathbf{v}_3 = (3, -5)$$

Dependent or independent??

Does there exist three numbers, c_1, c_2, c_3 , not all zero such that

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + c_3\mathbf{v}_3 = (0, 0)$$

$$c_1(1, -1) + c_2(2, -3) + c_3(3, -5)$$

I.e, does the system of equations

$$c_1 + 2c_2 + 3c_3 = 0$$

$$-c_1 - 3c_2 - 5c_3 = 0$$

have nontrivial solutions?

$$\left(\begin{array}{ccc|c} 1 & 2 & 3 & 0 \\ -1 & 3 & -5 & 0 \end{array} \right)$$

3. Four vectors: v_1, v_2, v_3, v_4 in \mathbb{R}^3

$$v_1 = (1, -1, 2), \quad v_2 = (2, -3, 0),$$

$$v_3 = (-1, -2, 2), \quad v_4 = (0, 4, -3)$$

Are there 4 real numbers c_1, c_2, c_3, c_4 ,

NOT ALL ZERO, such that

$$c_1 v_1 + c_2 v_2 + c_3 v_3 + c_4 v_4 = (0, 0, 0)?$$

$$\left(\begin{array}{cccc|c} 1 & 2 & -1 & 0 & 0 \\ -1 & -3 & -2 & 4 & 0 \\ 2 & 0 & 2 & -3 & 0 \end{array} \right)$$

- **A homogeneous system with more unknowns than equations ALWAYS has infinitely many nontrivial solutions.**

- **Let v_1, v_2, \dots, v_k be a set of k vectors in \mathbb{R}^n . If $k > n$, then the set of vectors is (automatically) linearly dependent.**

In \mathbb{R}^3 :

$$\mathbf{v}_1 = (1, -1, 2), \quad \mathbf{v}_2 = (2, -3, 0),$$

$$\mathbf{v}_3 = (-1, -2, 2), \quad \mathbf{v}_4 = (0, 4, -3)$$

Dependent or independent??

(a) $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}_4\}$

(b) $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$

(c) $\{\mathbf{v}_1, \mathbf{v}_2\}$

$$(b) \quad \mathbf{v}_1 = (1, -1, 2), \quad \mathbf{v}_2 = (2, -3, 0), \\ \mathbf{v}_3 = (-1, -2, 2)$$

Does $c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + c_3\mathbf{v}_3 = \mathbf{0}$

have non-trivial solutions? That is,

does

$$c_1 + 2c_2 - c_3 = 0$$

$$-c_1 - 3c_2 - 2c_3 = 0$$

$$2c_1 + 2c_3 = 0$$

have non-trivial solutions?

Augmented matrix and row reduce:

$$\left(\begin{array}{ccc|c} 1 & 2 & -1 & 0 \\ -1 & -3 & -2 & 0 \\ 2 & 0 & 2 & 0 \end{array} \right)$$

NOTE: It's enough to row reduce:

$$\left(\begin{array}{ccc} 1 & 2 & -1 \\ -1 & -3 & -2 \\ 2 & 0 & 2 \end{array} \right)$$

since the numbers to the right of the bar will always be 0.

Or, calculate the determinant

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

- $\det \neq 0$ implies unique solution

$$c_1 = c_2 = c_3 = 0$$

and **independent**,

- $\det = 0$ implies infinitely many solutions and **dependent**

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

Note: When testing a set of vectors

$\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$ for independence/dependence:

1. If you use the definition

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_k\mathbf{v}_k = \mathbf{0}$$

Then the vectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$ appear as columns.

2. If you use the determinant, then you can write the vectors either as rows or columns.

From the previous example:

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

This follows from the fact that to evaluate a determinant you can expand across any row or down any column.

$$4. \quad \mathbf{v}_1 = (a, 1, -1), \quad \mathbf{v}_2 = (-1, 2a, 3),$$

$$\mathbf{v}_3 = (-2, a, 2), \quad \mathbf{v}_4 = (3a, -2, a)$$

For what values of a are the vectors linearly dependent?

$$5. \quad \mathbf{v}_1 = (a, 1, -1), \quad \mathbf{v}_2 = (-1, 2a, 3),$$

$$\mathbf{v}_3 = (-2, a, 2)$$

For what values of a are the vectors
 $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$ linearly dependent?

$$a c_1 - c_2 - 2c_3 = 0$$

$$c_1 + 2a c_2 + a c_3 = 0$$

$$-c_1 + 3c_2 + 2c_3 = 0$$

Augmented matrix and row reduce:

$$\left(\begin{array}{ccc|c} a & -1 & -2 & 0 \\ 1 & 2a & a & 0 \\ -1 & 3 & 2 & 0 \end{array} \right)$$

Or row reduce:

$$\begin{pmatrix} a & -1 & -2 \\ 1 & 2a & a \\ -1 & 3 & 2 \end{pmatrix}$$

Or calculate the determinant

$$\begin{vmatrix} a & -1 & -2 \\ 1 & 2a & a \\ -1 & 3 & 2 \end{vmatrix}$$

5. $\mathbf{v}_1 = (1, -1, 2, 1), \quad \mathbf{v}_2 = (3, 2, 0, -1)$

$\mathbf{v}_3 = (-1, -4, 4, 3), \quad \mathbf{v}_4 = (2, 3, -2, -2)$

a. $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}_4\}$

dependent or independent?

b. If dependent, what is the maximum number of independent vectors?

Row reduce

$$\begin{pmatrix} 1 & -1 & 2 & 1 \\ 3 & 2 & 0 & -1 \\ -1 & -4 & 4 & 3 \\ 2 & 3 & -2 & -2 \end{pmatrix}$$

Or, calculate the determinant.

$$\begin{vmatrix} 1 & -1 & 2 & 1 \\ 3 & 2 & 0 & -1 \\ -1 & -4 & 4 & 3 \\ 2 & 3 & -2 & -2 \end{vmatrix}$$

$$v_1 = (1, -1, 2, 1), \quad v_2 = (3, 2, 0, -1),$$

$$v_3 = (-1, -4, 4, 3), \quad v_4 = (2, 3, -2, -2)$$

$$\begin{pmatrix} 1 & -1 & 2 & 1 \\ 3 & 2 & 0 & -1 \\ -1 & 4 & 4 & 3 \\ 2 & 3 & -2 & -2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -1 & 2 & 1 \\ 0 & 5 & -6 & -4 \\ 0 & -5 & 6 & 4 \\ 0 & 5 & -6 & -4 \end{pmatrix} \rightarrow$$

$$\begin{pmatrix} 1 & -1 & 2 & 1 \\ 0 & 5 & -6 & -4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

Tests for independence/dependence

Let $\mathcal{S} = \mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$ be a set of vectors in \mathbb{R}^n .

Case 1: $k > n$: \mathcal{S} is linearly dependent.

Case 2: $k = n$:

1. Solve the system of equations

$$c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \cdots + c_k \mathbf{v}_k = \mathbf{0}.$$

If a unique solution:

$$c_1 = c_2 = \cdots = c_n = 0,$$

the vectors are **independent**.

If infinitely many solutions:

The vectors are **dependent**.

OR 2. Form the $n \times n$ matrix A whose rows are v_1, v_2, \dots, v_n and row reduce A :

if the reduced matrix has n nonzero rows, i.e., if the rank of A is n , then **independent**;

if the reduced matrix has one or more zero rows, then **dependent**.

OR 3. Calculate $\det A$:

If $\det A \neq 0$,

the vectors are **independent**.

If $\det A = 0$,

the vectors are **dependent**.

Note: If $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ is a linearly independent set of vectors in \mathbb{R}^n , then each vector in \mathbb{R}^n has

a unique representation as a linear
combination of v_1, v_2, \dots, v_n .

Case 3: $k < n$: Form the $k \times n$ matrix A whose rows are $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$

1. Row reduce A :

if the reduced matrix has k nonzero rows – independent;

one or more zero rows – dependent.

Equivalently, solve the system of equations

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \cdots + c_k\mathbf{v}_k = \mathbf{0}.$$

If unique solution: $c_1 = c_2 = \cdots = c_n = 0$, then **independent**; If infinitely many solutions, then **dependent**.

Linear Independence/Dependence & Row Operations

Example. $v_1 = (1, -2, 3),$

$v_2 = (2, -3, 1), v_3 = (3, -4, -1)$

Dependent or independent?

$$\begin{pmatrix} 1 & -2 & 3 \\ 2 & -3 & 1 \\ 3 & -4 & -1 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & -2 & 3 \\ 0 & 1 & -5 \\ 0 & 2 & -10 \end{pmatrix}$$

$$\begin{pmatrix} 1 & -2 & 3 \\ 0 & 1 & -5 \\ 0 & 0 & 0 \end{pmatrix}$$

That is, R_3 is a linear combination of R_1 and R_2 ; the vectors are linearly dependent; the vector equation

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + c_3\mathbf{v}_3 = \mathbf{0}$$

has infinitely many non-zero solutions.

The General Result: Given a set of vectors $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ in \mathbb{R}^n .

Form the matrix V with $\mathbf{v}_1, \mathbf{v}_2 \dots$ as

rows and row reduce.

If you get a row of 0's, the vectors are linearly dependent and at least one of the vectors is a linear combination of the other vectors.

(THIS IS WHY A SYSTEM WITH INFINITELY MANY SOLUTIONS IS CALLED **DEPENDENT**. See Chapter 5, Part 1, pg 55.)

If you get no rows of 0's, the vectors are linearly independent

Given an $m \times n$ matrix A . $A : \mathbb{R}^n \rightarrow$

\mathbb{R}^m is a linear transformation!

$$A[\mathbf{v} + \mathbf{w}] = A\mathbf{v} + A\mathbf{w} \quad \text{and} \quad A[\alpha \mathbf{v}] = \alpha A\mathbf{v}$$

Example:

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

is a linear transformation from \mathbb{R}^3

to \mathbb{R}^2 .

Section 5.9. Eigenvalues/Eigenvectors

Example: Set

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

A is a linear transformation from \mathbb{R}^3 to \mathbb{R}^3 .

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

$$\begin{pmatrix} 1 & -3 & 1 \\ -1 & 1 & 1 \\ 3 & -3 & -1 \end{pmatrix} \begin{pmatrix} 1 \\ -3 \\ 2 \end{pmatrix} = \begin{pmatrix} 12 \\ -2 \\ 10 \end{pmatrix}$$

$$\begin{pmatrix} 1 & -3 & 1 \\ -1 & 1 & 1 \\ 3 & -3 & -1 \end{pmatrix} \begin{pmatrix} 3 \\ 2 \\ -3 \end{pmatrix} = \begin{pmatrix} -6 \\ -4 \\ 6 \end{pmatrix} = -2 \begin{pmatrix} 3 \\ 2 \\ -3 \end{pmatrix}$$

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

Let A be an $n \times n$ matrix.

A number λ is an **eigenvalue** of A if there is a **non-zero vector** \mathbf{v} such that

$$A\mathbf{v} = \lambda\mathbf{v}$$

How to find eigenvalues: Suppose λ is a number and \mathbf{v} is a non-zero vector such that $A\mathbf{v} = \lambda\mathbf{v}$. Then

To find the eigenvalues of A , find the values of λ that satisfy

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

Example: Let $A = \begin{pmatrix} 2 & 3 \\ 1 & 4 \end{pmatrix}$

$$A - \lambda I = \begin{pmatrix} 2 - \lambda & 3 \\ 1 & 4 - \lambda \end{pmatrix}$$

$$\det(A - \lambda I) = \begin{vmatrix} 2 - \lambda & 3 \\ 1 & 4 - \lambda \end{vmatrix}$$

$$= (2 - \lambda)(4 - \lambda) - 3 = \lambda^2 - 6\lambda + 5 = 0$$

Eigenvalues: $\lambda_1 = 5$, $\lambda_2 = 1$

$$\text{Let } A = \begin{pmatrix} 1 & -3 & 1 \\ -1 & 1 & 1 \\ 3 & -3 & -1 \end{pmatrix}$$

$$\det(A - \lambda I) = \begin{vmatrix} 1 - \lambda & -3 & 1 \\ -1 & 1 - \lambda & 1 \\ 3 & -3 & -1 - \lambda \end{vmatrix}$$

$$= -\lambda^3 + \lambda^2 + 4\lambda - 4$$

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

$$-\lambda^3 + \lambda^2 + 4\lambda - 4 = 0$$

or $\lambda^3 - \lambda^2 - 4\lambda + 4 = 0$

Terminology:

- $\det(A - \lambda I)$ is a polynomial of degree n , called the **characteristic polynomial** of A .
- The zeros of the characteristic polynomial are the eigenvalues of A .
- The equation $\det(A - \lambda I) = 0$ is called the **characteristic equation** of A .

- A **non-zero vector** \mathbf{v} that satisfies

$$A\mathbf{v} = \lambda\mathbf{v}$$

is called an **eigenvector** corresponding to the eigenvalue λ .

Note: Eigenvectors are NOT unique; an eigenvalue has infinitely many eigenvectors.

Examples: 1. $A = \begin{pmatrix} 2 & 2 \\ 2 & -1 \end{pmatrix}$

Characteristic polynomial:

$$\det(A - \lambda I) = \begin{vmatrix} 2 - \lambda & 2 \\ 2 & -1 - \lambda \end{vmatrix}$$

$$= (2 - \lambda)(-1 - \lambda) - 4$$

$$= \lambda^2 - \lambda - 6$$

Characteristic equation:

$$\lambda^2 - \lambda - 6 = (\lambda - 3)(\lambda + 2) = 0$$

Eigenvalues: $\lambda_1 = 3, \quad \lambda_2 = -2$

Eigenvectors:

$$(A - \lambda I) = \begin{pmatrix} 2 - \lambda & 2 \\ 2 & -1 - \lambda \end{pmatrix}$$

$\lambda_1 = 3$: Solve

$$(A - 3I)\mathbf{x} = \begin{pmatrix} -1 & 2 \\ 2 & -4 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$\lambda_2 = -2$: Solve

$$(A - (-2)I)\mathbf{x} = \begin{pmatrix} 4 & 2 \\ 2 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} =$$

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

Characteristic polynomial:

$$\begin{aligned} \det(A - \lambda I) &= \begin{vmatrix} 4 - \lambda & -1 \\ 4 & -\lambda \end{vmatrix} \\ &= \lambda^2 - 4\lambda + 4 \end{aligned}$$

Characteristic equation:

$$\lambda^2 - 4\lambda + 4 = (\lambda - 2)^2 = 0$$

Eigenvalues: $\lambda_1 = \lambda_2 = 2$

Eigenvectors:

$$(A - \lambda I) = \begin{pmatrix} 4 - \lambda & -1 \\ 4 & -\lambda \end{pmatrix}$$

$$\lambda_1 = \lambda_2 = 2$$

Solve:

$$(A - 2I)\mathbf{x} = \begin{pmatrix} 2 & -1 \\ 4 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$3. \quad A = \begin{pmatrix} 5 & 3 \\ -6 & -1 \end{pmatrix}$$

Characteristic polynomial:

$$\begin{aligned} \det(A - \lambda I) &= \begin{vmatrix} 5 - \lambda & 3 \\ -6 & -1 - \lambda \end{vmatrix} \\ &= \lambda^2 - 4\lambda + 13 \end{aligned}$$

Characteristic equation:

$$\lambda^2 - 4\lambda + 13 = 0$$

Eigenvalues:

$$\lambda_1 = 2 + 3i, \quad \lambda_2 = 2 - 3i$$

$$\text{Eigenvectors: } (A - \lambda I) = \begin{pmatrix} 5 - \lambda & 3 \\ -6 & -1 - \lambda \end{pmatrix}$$

$\lambda_1 = 2 + 3i$: Solve

$$(A - (2 + 3i)I)\mathbf{x} = \begin{pmatrix} 3 - 3i & 3 \\ -6 & -3 - 3i \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

NOTE: If $a + bi$ is an eigenvalue of A with eigenvector $\alpha + \beta i$, then $a - bi$ is also an eigenvalue of A and $\alpha - \beta i$ is a corresponding eigenvector.

$$4. \quad A = \begin{pmatrix} 4 & -3 & 5 \\ -1 & 2 & -1 \\ -1 & 3 & -2 \end{pmatrix}$$

Characteristic polynomial:

$$\begin{aligned} \det(A - \lambda I) &= \begin{vmatrix} 4 - \lambda & -3 & 5 \\ -1 & 2 - \lambda & -1 \\ -1 & 3 & -2 - \lambda \end{vmatrix} \\ &= -\lambda^3 + 4\lambda^2 - \lambda - 6 \end{aligned}$$

Characteristic equation:

$$\lambda^3 - 4\lambda^2 + \lambda + 6 = (\lambda - 3)(\lambda - 2)(\lambda + 1) = 0.$$

Eigenvalues: $\lambda_1 = 3$, $\lambda_2 = 2$, $\lambda_3 =$
 -1

Eigenvectors:

$$(A - \lambda I) = \begin{pmatrix} 4 - \lambda & -3 & 5 \\ -1 & 2 - \lambda & -1 \\ -1 & 3 & -2 - \lambda \end{pmatrix}$$

$$\lambda_1 = 3:$$

$$(A - \lambda I) = \begin{pmatrix} 4 - \lambda & -3 & 5 \\ -1 & 2 - \lambda & -1 \\ -1 & 3 & -2 - \lambda \end{pmatrix}$$

$$\lambda_2 = 2$$

$$(A - \lambda I) = \begin{pmatrix} 4 - \lambda & -3 & 5 \\ -1 & 2 - \lambda & -1 \\ -1 & 3 & -2 - \lambda \end{pmatrix}$$

$$\lambda_3 = -1$$

$$5. \quad A = \begin{pmatrix} 4 & 1 & -1 \\ 2 & 5 & -2 \\ 1 & 1 & 2 \end{pmatrix}$$

Characteristic polynomial:

$$\begin{aligned} \det(A - \lambda I) &= \begin{vmatrix} 4 - \lambda & 1 & -1 \\ 2 & 5 - \lambda & -2 \\ 1 & 1 & 2 - \lambda \end{vmatrix} \\ &= -\lambda^3 + 11\lambda^2 - 39\lambda + 45 \end{aligned}$$

Characteristic equation:

$$\lambda^3 - 11\lambda^2 + 39\lambda - 45 = (\lambda - 3)^2(\lambda - 5) = 0.$$

Eigenvalues: $\lambda_1 = 5, \lambda_2 = \lambda_3 = 3$

Eigenvectors:

$$(A - \lambda I) = \begin{pmatrix} 4 - \lambda & 1 & -1 \\ 2 & 5 - \lambda & -2 \\ 1 & 1 & 2 - \lambda \end{pmatrix}$$

$$\lambda_1 = 5$$

$$(A - \lambda I) = \begin{pmatrix} 4 - \lambda & 1 & -1 \\ 2 & 5 - \lambda & -2 \\ 1 & 1 & 2 - \lambda \end{pmatrix}$$

$$\lambda_2 = \lambda_3 = 3$$

$$6. \quad A = \begin{pmatrix} -3 & 1 & -1 \\ -7 & 5 & -1 \\ -6 & 6 & -2 \end{pmatrix}$$

Characteristic polynomial:

$$\begin{aligned} \det(A - \lambda I) &= \begin{vmatrix} -3 - \lambda & 1 & -1 \\ -7 & 5 - \lambda & -1 \\ -6 & 6 & -2 - \lambda \end{vmatrix} \\ &= -\lambda^3 + 12\lambda + 16 \end{aligned}$$

Characteristic equation:

$$\lambda^3 - 12\lambda - 16 = (\lambda - 4)(\lambda + 2)^2 = 0.$$

Eigenvalues: $\lambda_1 = 4, \lambda_2 = \lambda_3 = -2$

Eigenvectors:

$$(A - \lambda I) = \begin{pmatrix} -3 - \lambda & 1 & -1 \\ -7 & 5 - \lambda & -1 \\ -6 & 6 & -2 - \lambda \end{pmatrix}$$

$$\lambda_1 = 4:$$

$$(A - \lambda I) = \begin{pmatrix} -3 - \lambda & 1 & -1 \\ -7 & 5 - \lambda & -1 \\ -6 & 6 & -2 - \lambda \end{pmatrix}$$

$$\lambda_2 = \lambda_3 = -2:$$

Some Facts

THEOREM If

$$\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$$

are eigenvectors of a matrix A corresponding to distinct eigenvalues

$$\lambda_1, \lambda_2, \dots, \lambda_k,$$

then $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$ are linearly independent.

THEOREM Let A be a (real) $n \times n$ matrix. If the complex number $\lambda = a + bi$ is an eigenvalue of A with corresponding (complex) eigenvector $\mathbf{u} + i\mathbf{v}$, then $\lambda = a - bi$, the *conjugate* of $a + bi$, is also an eigenvalue of A and $\mathbf{u} - i\mathbf{v}$ is a corresponding eigenvector.

THEOREM Let A be an $n \times n$ matrix with eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$.

Then

$$\det A = (-1)^n \lambda_1 \cdot \lambda_2 \cdot \lambda_3 \cdot \dots \cdot \lambda_n.$$

That is, $\det A$ is the ± 1 times the product of the eigenvalues of A .

(The λ 's are not necessarily distinct, the multiplicity of an eigenvalue may be greater than 1, and they are not necessarily real.)

Equivalences: (A an $n \times n$ matrix)

1. $A\mathbf{x} = \mathbf{b}$ has a unique solution.
2. The reduced row echelon form of A is I_n .
3. The rank of A is n .
4. A has an inverse.
5. $\det A \neq 0$.
6. 0 is not an eigenvalue of A .

The following are equivalent:

1. $\det A = 0$.
2. A does not have an inverse.
3. The rank of A is less than n .
4. The reduced row echelon form of A is not I_n .
5. The system $A\mathbf{x} = \mathbf{b}$ does not have a unique solution.
6. 0 is an eigenvalue of A .