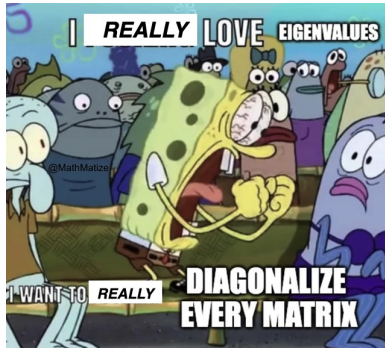


MATH 200C: Linear Algebra



Class 27: Monday, April 20, 2026



- ▶ Diagonalization
- ▶ Notes on Assignment 24

Eigenvalues and Eigenvectors

Definition: An **eigenvector** of an $n \times n$ matrix A is a nonzero vector \mathbf{x} such that $A\mathbf{x} = \lambda\mathbf{x}$ for some scalar λ .

A scalar λ is called an **eigenvalue** of A if there is a nontrivial solution \mathbf{x} of $A\mathbf{x} = \lambda\mathbf{x}$; such an \mathbf{x} is called an eigenvector corresponding to λ .

Theorem 2: If $\mathbf{v}_1, \dots, \mathbf{v}_r$ are eigenvectors that correspond to distinct eigenvalues $\lambda_1, \dots, \lambda_r$ of an $n \times n$ matrix A , then the set $\{\mathbf{v}_1, \dots, \mathbf{v}_r\}$ is linearly independent.

From Last Time:

Suppose $\lambda_1, \lambda_2, \lambda_3$ are distinct eigenvalues of a square matrix A
with associated eigenvectors $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$

Thus $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$ are nonzero vectors

$$A\mathbf{v}_1 = \lambda_1\mathbf{v}_1, A\mathbf{v}_2 = \lambda_2\mathbf{v}_2, A\mathbf{v}_3 = \lambda_3\mathbf{v}_3$$

$$\lambda_1 \neq \lambda_2, \lambda_1 \neq \lambda_3, \lambda_2 \neq \lambda_3$$

Suppose $\{\mathbf{v}_1, \mathbf{v}_2\}$ is linearly independent, but $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ is linearly dependent.

Then (*) $\mathbf{v}_3 = a_1\mathbf{v}_1 + a_2\mathbf{v}_2$ for some constants a_1, a_2

Multiply (*) by λ_3 to obtain

lan Equation: $\lambda_3\mathbf{v}_3 = a_1\lambda_3\mathbf{v}_1 + a_2\lambda_3\mathbf{v}_2$

Multiply (*) by A to obtain **Lambda Equation**

$$A\mathbf{v}_3 = A(a_1\mathbf{v}_1 + a_2\mathbf{v}_2) = Aa_1\mathbf{v}_1 + Aa_2\mathbf{v}_2 = a_1\lambda_1\mathbf{v}_1 + a_2\lambda_2\mathbf{v}_2$$

$$A\mathbf{v}_3 = a_1\lambda_1\mathbf{v}_1 + a_2\lambda_2\mathbf{v}_2$$

Left Hand side equals $\lambda_3\mathbf{v}_3$ so **Lambda Equation** is now

$$\lambda_3\mathbf{v}_3 = a_1\lambda_1\mathbf{v}_1 + a_2\lambda_2\mathbf{v}_2$$

$$\text{Recall (*) } \mathbf{v}_3 = a_1\mathbf{v}_1 + a_2\mathbf{v}_2$$

Ian Equation: $\lambda_3\mathbf{v}_3 = a_1\lambda_3\mathbf{v}_1 + a_2\lambda_3\mathbf{v}_2$

Lambda Equation: $\lambda_3\mathbf{v}_3 = a_1\lambda_1\mathbf{v}_1 + a_2\lambda_2\mathbf{v}_2$

Subtract **Lambda Equation** From **Ian Equation:**

$$\mathbf{0} = a_1(\lambda_3 - \lambda_1)\mathbf{v}_1 + a_2(\lambda_3 - \lambda_2)\mathbf{v}_2$$

But $\{\mathbf{v}_1, \mathbf{v}_2\}$ is linearly independent

$$\text{so } a_1(\lambda_3 - \lambda_1) = 0 \text{ and } a_2(\lambda_3 - \lambda_2) = 0$$

Since $(\lambda_3 - \lambda_1) \neq 0$ and $(\lambda_3 - \lambda_2) \neq 0$, we must have

$$a_1 = 0, a_2 = 0$$

which makes (*) $\mathbf{v}_3 = 0\mathbf{v}_1 + 0\mathbf{v}_2 = \mathbf{0}$ which contradicts $\mathbf{v}_3 \neq \mathbf{0}$

Proof in General Case

Suppose $\lambda_1, \lambda_2, \dots, \lambda_p, \lambda_{p+1}$ are distinct eigenvalues of a square matrix A with associated eigenvectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p, \mathbf{v}_{p+1}$,

Thus $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p, \mathbf{v}_{p+1}$, are all nonzero vectors

$$A\mathbf{v}_i = \lambda_i\mathbf{v}_i \text{ for } i = 1, 2, \dots, p, p+1,$$

$$\lambda_i \neq \lambda_j \text{ if } i \neq j$$

Suppose $\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$ is linearly independent, but $\{\mathbf{v}_1, \dots, \mathbf{v}_p, \mathbf{v}_{p+1}\}$ is linearly dependent.

Then (*) $\mathbf{v}_{p+1} = a_1\mathbf{v}_1 + a_2\mathbf{v}_2 + \dots + a_p\mathbf{v}_p$ for some constants a_1, a_2, \dots, a_p

Multiply (*) by λ_{p+1} to obtain

$$\text{lan Equation: } \lambda_{p+1}\mathbf{v}_{p+1} = a_1\lambda_{p+1}\mathbf{v}_1 + a_2\lambda_{p+1}\mathbf{v}_2 + \dots + a_p\lambda_{p+1}\mathbf{v}_p$$

Multiply (*) by A to obtain **Lambda Equation**

$$A\mathbf{v}_{p+1} = A(a_1\mathbf{v}_1 + a_2\mathbf{v}_2 + \dots + a_p\mathbf{v}_p) = \\ Aa_1\mathbf{v}_1 + Aa_2\mathbf{v}_2 + \dots + Aa_p\mathbf{v}_p = a_1\lambda_1\mathbf{v}_1 + a_2\lambda_2\mathbf{v}_2 + \dots + a_p\lambda_p\mathbf{v}_p$$

$$A\mathbf{v}_{p+1} = a_1\lambda_1\mathbf{v}_1 + a_2\lambda_2\mathbf{v}_2 + \cdots + a_p\lambda_p\mathbf{v}_p$$

Left Hand side equals $\lambda_{p+1}\mathbf{v}_{p+1}$ so **Lambda Equation** is now

$$\lambda_{p+1}\mathbf{v}_{p+1} = a_1\lambda_1\mathbf{v}_1 + a_2\lambda_2\mathbf{v}_2 + \cdots + a_p\lambda_p\mathbf{v}_p$$

Recall (*) $\mathbf{v}_{p+1} = a_1\mathbf{v}_1 + a_2\mathbf{v}_2 + \cdots + a_p\mathbf{v}_p$ for some constants

$$a_1, a_2, \dots, a_p$$

Ian Equation: $\lambda_{p+1}\mathbf{v}_{p+1} = a_1\lambda_{p+1}\mathbf{v}_1 + a_2\lambda_{p+1}\mathbf{v}_2 + \cdots + a_p\lambda_{p+1}\mathbf{v}_p$

Lambda Equation: $\lambda_{p+1}\mathbf{v}_{p+1} = a_1\lambda_1\mathbf{v}_1 + a_2\lambda_2\mathbf{v}_2 + \cdots + a_p\lambda_p\mathbf{v}_p$

Subtract **Lambda Equation** From **Ian Equation**:

$$\mathbf{0} = a_1(\lambda_{p+1} - \lambda_1)\mathbf{v}_1 + a_2(\lambda_{p+1} - \lambda_2)\mathbf{v}_2 + \cdots + a_p(\lambda_{p+1} - \lambda_p)\mathbf{v}_p$$

But $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$ is linearly independent

so $a_i(\lambda_{p+1} - \lambda_i) = 0$ for $i = 1, 2, \dots, p$

Since $(\lambda_{p+1} - \lambda_i) \neq 0$ for $i = 1, 2, \dots, p$

we must have $a_i = 0$ for $i = 1, 2, \dots, p$

which makes (*) $\mathbf{v}_{p+1} = 0\mathbf{v}_1 + 0\mathbf{v}_2 + \cdots + 0\mathbf{v}_p = \mathbf{0}$ which

contradicts $\mathbf{v}_{p+1} \neq \mathbf{0}$

Similar Matrices

Definition: Two $n \times n$ square matrices A and B are **similar** if there is an $n \times n$ invertible matrix P such that $P^{-1}AP = B$.

Theorem: If $n \times n$ square matrices A and B are similar, then they have the same characteristic polynomials and thus the same eigenvalues with the same multiplicities.

Diagonalization

The eigenvalue–eigenvector information contained within a matrix A often can be displayed in a useful factorization of the form $A = PDP^{-1}$ where D is a diagonal matrix.

The factorization enables us to compute A^k quickly for large values of k , a fundamental idea in several applications of linear algebra.

Theorem 5: The Diagonalization Theorem An $n \times n$ matrix A is diagonalizable if and only if A has n linearly independent eigenvectors.

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In fact, $A = PDP^{-1}$, with D a diagonal matrix, if and only if the columns of P are n linearly independent eigenvectors of A . In this case, the diagonal entries of D are eigenvalues of A that correspond, respectively, to the eigenvectors in P .

Theorem 6: An $n \times n$ matrix with n distinct eigenvalues is diagonalizable.

Theorem 7 Let A be an $n \times n$ matrix whose distinct eigenvalues are $\lambda_1, \dots, \lambda_p$

- ▶ (a) For $1 \leq k \leq p$, the dimension of the eigenspace for λ_k is less than or equal to the multiplicity of the eigenvalue λ_k .
- ▶ (b) The matrix A is diagonalizable if and only if the sum of the dimensions of the eigenspaces equals n , and this happens if and only if
 - ▶ (i) the characteristic polynomial factors completely into linear factors and
 - ▶ (ii) the dimension of the eigenspace for each λ_k equals the algebraic multiplicity of λ_k .
- ▶ (c) If A is diagonalizable and B_k is a basis for the eigenspace corresponding to λ_k for each k , then the total collection of vectors in the sets B_1, \dots, B_p forms an eigenvector basis for R^n .