

Computer Vision

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Textbook

□ **Linear Algebra and its application**
by David C Lay

Reference books

□ **Elementary Linear Algebra**
by Howard Anton and Chris Rorres

References

Readings for these lecture notes:

**Linear Algebra and its application
by David C Lay**

These notes contain material from the above
recourses.

The Invertible Matrix Theorem (Continued)

Theorem: Let A be an $n \times n$ matrix. Then A is invertible if and only if:

- s. The number 0 is *not* an eigenvalue of A .
- t. The determinant of A is *not* zero.

○ In linear algebra, **eigenvalues** and **eigenvectors** are used to understand the behavior of a linear transformation represented by a square matrix.

○ **Eigenvector**: A **non-zero vector** \vec{v} such that when a matrix **A** is multiplied by \vec{v} , the result is a **scalar multiple** of \vec{v}

$$A\vec{v} = \alpha\vec{v}$$

Eigenvalue λ : The scalar in the equation above. It tells us how the direction of the vector is scaled.

○ To find **eigenvalues**: Solve **$\det(A - \lambda I) = 0$**

○ To find **eigenvectors**: Solve **$(A - \lambda I)v = 0$**

Example

$A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$. Find eigenvalues and eigenvectors.

Solution:

Eigenvalues: Solve $\det(A - \lambda I) = 0$

$$\det(A - \lambda I) = \det\left(\begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}\right)$$

$$= \det\left(\begin{bmatrix} 2 - \lambda & 1 \\ 1 & 2 - \lambda \end{bmatrix}\right)$$

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

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

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

$$\lambda = 3 \text{ and } \lambda = 1$$

Eigenvectors: Solve $(A - \lambda I)v = 0$

For $\lambda = 1$

$$(A - I)v = 0$$

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

$$A - I = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$$

Row reduce the augmented matrix for $(A - I)v = 0$

$$\begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \end{bmatrix}$$

$$-1 \times R_1 + R_2 \rightarrow \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\Rightarrow x_1 + x_2 = 0$$

$$\Rightarrow x_1 = -x_2$$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -x_2 \\ x_2 \end{bmatrix} = x_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$\text{Eigenvector} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

Example 4: Let $A = \begin{bmatrix} 4 & -1 & 6 \\ 2 & 1 & 6 \\ 2 & -1 & 8 \end{bmatrix}$. An **eigenvalue** of **A** is **2**. Find a **basis** for the corresponding **eigenspace**.

Solution:

$$\begin{aligned} \mathbf{A} - 2\mathbf{I} &= \begin{bmatrix} 4 & -1 & 6 \\ 2 & 1 & 6 \\ 2 & -1 & 8 \end{bmatrix} - \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} \\ &= \begin{bmatrix} 2 & -1 & 6 \\ 2 & -1 & 6 \\ 2 & -1 & 6 \end{bmatrix} \end{aligned}$$

and row reduce the augmented matrix for $(\mathbf{A} - \mathbf{I})\mathbf{x} = \mathbf{0}$

$$[A - 2I \ 0] = \begin{bmatrix} 2 & -1 & 6 & 0 \\ 2 & -1 & 6 & 0 \\ 2 & -1 & 6 & 0 \end{bmatrix}$$

$$-1 \times R_1 + R_2, -1 \times R_1 + R_3 \longrightarrow \begin{bmatrix} 2 & -1 & 6 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\Rightarrow 2x_1 - x_2 + 6x_3 = 0$$

The general solution is $x_1 = \frac{1}{2}x_2 - 3x_3$ with x_2 and x_3 are **free variables**.

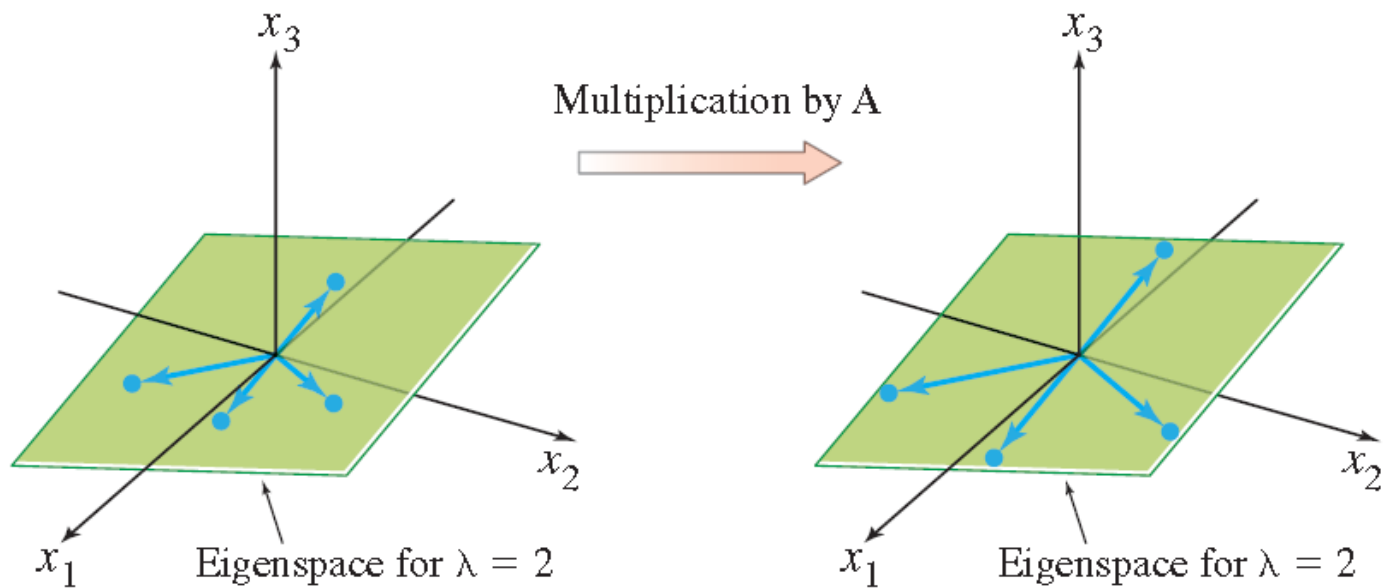
$x_1 = \frac{1}{2}x_2 - 3x_3$ with x_2 and x_3 are **free variables**

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

The **eigenspace**, shown in the following figure, is a two-dimensional subspace of \mathbb{R}^3 . A **basis** is

$$\left\{ \begin{bmatrix} \frac{1}{2} \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix} \right\}$$

The **eigenspace**, shown in the figure below, is a two-dimensional subspace of \mathbb{R}^3



A acts as a dilation on the eigenspace.

Eigenvectors And Eigenvalues

Theorem 1: The **eigenvalues** of a **triangular matrix** are the **entries** on its **main diagonal**.

Proof: For simplicity, consider the **3×3** case.

If **A** is **upper triangular**, the **$A - \lambda I$** has the form

$$\begin{aligned} A - \lambda I &= \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22} & a_{23} \\ 0 & 0 & a_{33} \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} a_{11} - \lambda & a_{12} & a_{13} \\ 0 & a_{22} - \lambda & a_{23} \\ 0 & 0 & a_{33} - \lambda \end{bmatrix} \end{aligned}$$

□ The scalar λ is an eigenvalue of A if and only if the equation $(A - \lambda I)\mathbf{x} = \mathbf{0}$ has a **nontrivial solution**, that is, if and only if the equation has a **free variable**.

□ Because of the **zero entries** in $A - \lambda I$, it is easy to see that $(A - \lambda I)\mathbf{x} = \mathbf{0}$ has a **free variable** if and only if **at least one of the entries** on the **diagonal** of $A - \lambda I$ is zero.

□ This happens if and only if λ equals one of the entries a_{11}, a_{22}, a_{33} in A .

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

Example 5

$$\text{Let } A = \begin{bmatrix} \mathbf{3} & 6 & -8 \\ 0 & \mathbf{0} & 6 \\ 0 & 0 & \mathbf{2} \end{bmatrix} \text{ and } B = \begin{bmatrix} \mathbf{4} & 0 & 0 \\ -2 & \mathbf{1} & 0 \\ 5 & 3 & \mathbf{4} \end{bmatrix}.$$

The **eigenvalues** of **A** are **3**, **0**, and **2**.

The **eigenvalues** of **B** are **4** and **1**.

What does it mean for a matrix **A** to have an **eigenvalue** of **0**, such as in Example 5?

Solution

This happens if and only if the equation

$$A\mathbf{x} = \mathbf{0} \text{ -----(1)}$$

has a **nontrivial solution**. But (1) is equivalent to

$A\mathbf{x} = \mathbf{0}$, which has a **nontrivial solution** if and only if A is **not invertible**.

□ Thus **0** is an **eigenvalue** of A *if and only if* A is **not invertible**.

Note: This fact will be added to the **Invertible Matrix Theorem in Section**

Eigenvectors and Eigenvalues

Theorem 2: If A is a **triangular matrix**, then **$\det A$** is the **product of the entries** on the **main diagonal** of **A** .

Determinants

- Let A be an $n \times n$ matrix, let U be any echelon form obtained from A by **row replacements** and **row interchanges (without scaling)**, and let r be the number of such **row interchanges**.
- Then the **determinant** of A , written as **$\det A$** , is **$(-1)^r$** times the **product** of the **diagonal entries** u_{11}, \dots, u_{nn} in U .
- If A is **invertible**, then u_{11}, \dots, u_{nn} are **all pivots** (because $A \sim I_n$ and the u_{ij} have not been **scaled** to 1's).

Otherwise, at least u_{nn} is **zero**, and the product $u_{11} \dots u_{nn}$ is **zero**.

If there are r interchanges

$$\begin{cases} (-1)^r \cdot (\text{product of pivots in U}) & \text{when A is invertible} \\ 0 & \text{when A is not invertible} \end{cases} \quad \text{---(1)}$$

Example 1: Compute $\det A$ for

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

Solution: The following row reduction uses one row interchange

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

$$-2 \times R_1 + R_2 \rightarrow \begin{bmatrix} 1 & 5 & 0 \\ 0 & -6 & -1 \\ 0 & -2 & 0 \end{bmatrix}$$

$$R_2 \leftrightarrow R_3 \rightarrow \begin{bmatrix} 1 & 5 & 0 \\ 0 & -2 & 0 \\ 0 & -6 & -1 \end{bmatrix} \quad (r = 1)$$

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

$$-3 \times R_2 + R_3 \rightarrow \begin{bmatrix} \mathbf{1} & 5 & 0 \\ 0 & \mathbf{-2} & 0 \\ 0 & 0 & \mathbf{-1} \end{bmatrix} = \mathbf{U}_1$$

$$\begin{aligned} \det A &= (-1)^1(1)(-2)(-1) \\ &= -2 \end{aligned}$$

Alternative Solution

□ The following alternative row reduction avoids the **row interchange** and produces a **different echelon form**.

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

$$2 \times R_1 + R_2 \rightarrow \begin{bmatrix} 1 & 5 & 0 \\ 0 & -6 & -1 \\ 0 & -2 & 0 \end{bmatrix}$$

$$-\frac{1}{3} \times R_2 + R_3 \rightarrow \begin{bmatrix} 1 & 5 & 0 \\ 0 & -6 & -1 \\ 0 & 0 & \frac{1}{3} \end{bmatrix} = U_2$$

$$\det A = (-1)^0(1)(-6)\left(\frac{1}{3}\right)$$

= -2. The answer is same as before.

The Invertible Matrix Theorem (Continued)

Theorem: Let A be an $n \times n$ matrix. Then A is invertible if and only if:

- s. The number 0 is *not* an eigenvalue of A .
- t. The determinant of A is *not* zero.

Properties of Determinants

Theorem 3: Let A and B be $n \times n$ matrices.

- a. A is invertible if and only if $\det \neq 0$
- b. $\det \mathbf{AB} = (\det \mathbf{A})(\det \mathbf{B})$
- c. $\det \mathbf{A}^T = \det \mathbf{A}$

Properties of Determinants

- d. If A is **triangular**, then $\det A$ is the **product of the entries** on the **main diagonal** of A .
- e. A **row replacement operation** on A does not change the **determinant**. A **row interchange** changes the **sign** of the determinant. A **row scaling** also **scales** the determinant by the **same scalar factor**.

The Characteristic Equation

Theorem 3(a) shows how to determine when a matrix of the form $(A - \lambda I)x$ is **not invertible**.

- The **scalar equation** $\det (A - \lambda I)x = 0$ is called the **characteristic equation** of A .
- A scalar λ is an **eigenvalue** of an $n \times n$ **matrix A** if and only if λ satisfies the characteristic equation
 $\det(A - \lambda I)x = 0$

Example 3: Find the **characteristic equation** of

$$A = \begin{bmatrix} 5 & -2 & 6 & -1 \\ 0 & 3 & -8 & 0 \\ 0 & 0 & 5 & 4 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Solution: Form $A - \lambda I$, and use Theorem 3(d):

$$\det (A - \lambda I)$$

$$= \det \begin{bmatrix} 5 - \lambda & -2 & 6 & -1 \\ 0 & 3 - \lambda & -8 & 0 \\ 0 & 0 & 5 - \lambda & 4 \\ 0 & 0 & 0 & 1 - \lambda \end{bmatrix}$$

$$= (5 - \lambda) (3 - \lambda) (5 - \lambda) (1 - \lambda)$$

The **characteristic equation** is

$$(5 - \lambda)^2 (3 - \lambda)(1 - \lambda) = 0$$

$$\Rightarrow (5 - \lambda)^2 (\lambda - 3)(\lambda - 1) = 0$$

Expanding the product, we can also write

$$\lambda^4 - 14\lambda^3 + 68\lambda^2 - 130\lambda + 75 = 0$$

$$\lambda^4 - 14\lambda^3 + 68\lambda^2 - 130\lambda + 75 = 0$$

If A is an $n \times n$ matrix, then $\det (A - \lambda I)$ is a polynomial of degree n called the **characteristic polynomial** of A .

- The **eigenvalue 5** in Example 3 is said to have ***multiplicity 2*** because $(\lambda - 5)$ occurs **two times** as a factor of the **characteristic polynomial**.
- In general, the **(algebraic) multiplicity** of an **eigenvalue λ** is its multiplicity as a **root** of the **characteristic equation**.

Example 4 The characteristic polynomial of a 6×6 matrix is $\lambda^6 - 4\lambda^5 - 12\lambda^4$. Find the **eigenvalues** and their **multiplicities**.

Factor the polynomial

$$\begin{aligned}\lambda^6 - 4\lambda^5 - 12\lambda^4 &= \lambda^4(\lambda^2 - 4\lambda - 12) = \lambda^4(\lambda^2 - 6\lambda + 2\lambda - 12) \\ &= \lambda^4(\lambda - 6)(\lambda + 2)\end{aligned}$$

□ The **eigenvalues** are **0 (multiplicity 4)**, **6 (multiplicity 1)**, and **-2 (multiplicity 1)**. We could also list the **eigenvalues** in Example 4 as **0, 0, 0, 0, 6**, and **-2**, so that the **eigenvalues** are **repeated** according to their **multiplicities**.

Orthogonal Matrix

Orthogonal Matrix: A square invertible matrix U such that $U^{-1} = U^T$.

Theorem 6 An $m \times n$ matrix U has orthonormal columns if and only if $U^T U = I$.

The Singular Value Decomposition

The **decomposition** of **A** involves an **m × n** “diagonal” matrix **Σ** of the form

$$\Sigma = \begin{bmatrix} D & 0 \\ 0 & 0 \end{bmatrix} \quad \leftarrow \boxed{\text{m - r rows}} \quad \text{-----} (1)$$

↑

$\boxed{\text{n - r columns}}$

$$A_{m \times n} = U_{m \times m} \Sigma_{m \times n} V^T_{n \times n}$$

□ where **D** is an **r × r diagonal matrix** for **some r not exceeding the smaller of m and n**. (If r equals m or n or both, some or all of the zero matrices do not appear.)

The Singular Value Decomposition

Theorem 10: The Singular Value Decomposition Let A be an $m \times n$ matrix with rank r . Then there exists an $m \times n$ matrix Σ as in (1) for which the diagonal entries in D are the first r singular values of A , $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > 0$, and there exist an $m \times m$ orthogonal matrix U and an $n \times n$ orthogonal matrix V such that

$$A = U\Sigma V^T$$
$$A_{m \times n} = U_{m \times m} \Sigma_{m \times n} V_{n \times n}^T$$

The Singular Value Decomposition

- Any factorization $A = U\Sigma V^T$, with U and V orthogonal, Σ as in (1), and positive diagonal entries in D , is called a **singular value decomposition** (or **SVD**) of A .
- The **columns** of U in such a decomposition are called **left singular vectors** of A , and the **columns** of V are called **right singular vectors** of A .