

# **Mathematics Fundamentals**

## Lecture 5: More on Matrices

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# Overview of Lecture 5

Determinants

Matrix Inverses

Rank of a Matrix

Matrix Decompositions

## Determinants

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# Introduction to Determinants

**Determinants** are a scalar value that can be computed from the elements of a square matrix. They provide key information about the matrix, including whether it is invertible.

- Determinants have important applications in areas such as linear algebra, geometry, and differential equations.
- Geometric interpretation: The determinant of a  $2 \times 2$  matrix can be interpreted as the area scaling factor for linear transformations.
- If the determinant of a matrix is zero, the matrix is **singular**, meaning it does not have an inverse.

# Calculating Determinants

**2x2 Matrix:**

$$\det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = ad - bc$$

**3x3 Matrix:**  $\det(A) = aei + bfg + cdh - ceg - bdi - afh$

**4x4 Matrix:**

$$\begin{aligned}\det(A) = & a(f(kp - lo) - g(jp - ln) + h(jo - kn)) \\ & - b(e(kp - lo) - g(ip - mn) + h(io - km)) \\ & + c(e(jp - ln) - f(ip - mn) + h(io - jm)) \\ & - d(e(jo - kn) - f(io - km) + g(io - jm))\end{aligned}$$

- Determinants become increasingly complex to compute as the size of the matrix increases, often requiring advanced techniques like cofactor expansion or matrix decomposition.

# Properties of Determinants

Determinants have several important properties:

- **Product Rule:**  $\det(AB) = \det(A) \times \det(B)$
- **Transpose:**  $\det(A^T) = \det(A)$
- **Identity Matrix:**  $\det(I) = 1$  for any identity matrix  $I$
- **Triangular Matrix:** The determinant of a triangular matrix (upper or lower) is the product of its diagonal elements:  
$$\det(A) = a_{11} \times a_{22} \times \cdots \times a_{nn}$$
- **Zero Row/Column:** If a matrix has a row or a column consisting entirely of zeros, then  $\det(A) = 0$

## Matrix Inverses

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## Introduction to Matrix Inverses

A square matrix  $A$  has an **inverse**  $A^{-1}$  if and only if:

$$AA^{-1} = A^{-1}A = I$$

Where  $I$  is the identity matrix. Inverses have several key properties:

- A matrix is invertible only if its determinant is non-zero.
- $(AB)^{-1} = B^{-1}A^{-1}$
- $(A^T)^{-1} = (A^{-1})^T$
- $(A^{-1})^{-1} = A$

## Connection Between Inverses and Determinants

For a  $2 \times 2$  matrix  $A$ , the inverse is given by:

$$A^{-1} = \frac{1}{\det(A)} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

- The determinant plays a crucial role in determining whether an inverse exists.
- For larger matrices, the calculation of the inverse relies on methods like Gaussian elimination or LU decomposition.

## Rank of a Matrix

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# Linear Independence

- A set of vectors  $\{v_1, v_2, \dots, v_n\}$  is linearly independent if the only solution to the equation:

$$c_1 v_1 + c_2 v_2 + \cdots + c_n v_n = 0$$

is  $c_1 = c_2 = \cdots = c_n = 0$ .

- If any of the coefficients  $c_i$  can be non-zero while still satisfying the equation, the vectors are linearly dependent.

## Example:

- Vectors  $v_1 = \begin{pmatrix} 1 & 0 \end{pmatrix}$  and  $v_2 = \begin{pmatrix} 0 & 1 \end{pmatrix}$  are linearly independent because the only solution to  $c_1 v_1 + c_2 v_2 = 0$  is  $c_1 = 0$  and  $c_2 = 0$ .
- Vectors  $v_1 = \begin{pmatrix} 1 & 2 \end{pmatrix}$  and  $v_2 = \begin{pmatrix} 2 & 4 \end{pmatrix}$  are linearly dependent because  $v_2 = 2v_1$ , meaning  $c_1 = 2$  and  $c_2 = -1$  is a non-trivial solution to  $c_1 v_1 + c_2 v_2 = 0$ .

## Rank of a Matrix

The **rank** of a matrix is the maximum number of linearly independent rows (or columns) in the matrix.

- A matrix is full-rank if all its rows (or columns) are linearly independent.

**Example:**

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$$

- The rows of  $A$  are linearly dependent because the third row can be written as a linear combination of the first two rows ( $R_3 = 2R_2 - R_1$ ).
- The rank of  $A$  is 2, as only two of the rows (or columns) are linearly independent.

# Matrix Decompositions

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# Matrix Decompositions

Matrix decompositions are methods to factorize a matrix into products of simpler matrices. These are useful for simplifying matrix operations.

- **LU Decomposition:**  $A = LU$  where  $L$  is lower triangular and  $U$  is upper triangular.
- **Cholesky Decomposition:** For positive definite matrices,  $A = LL^T$ .
- **QR Decomposition:**  $A = QR$ , where  $Q$  is orthogonal and  $R$  is upper triangular.
- Decompositions simplify calculations of determinants, inverses, and solutions to linear systems.

## Determinant Computation Using LU Decomposition

If a matrix  $A$  can be decomposed into an LU decomposition, such that  $A = LU$ , then the determinant of  $A$  can be computed as:

$$\det(A) = \det(L) \times \det(U)$$

Since the determinant of a triangular matrix is the product of its diagonal elements:

$$\det(L) = l_{11} \times l_{22} \times \cdots \times l_{nn}, \quad \det(U) = u_{11} \times u_{22} \times \cdots \times u_{nn}$$

Therefore:

$$\det(A) = \left( \prod_{i=1}^n l_{ii} \right) \times \left( \prod_{i=1}^n u_{ii} \right)$$

# Introduction to Spectral Decomposition

**Spectral decomposition**, also known as eigenvalue decomposition, is a method where a matrix is expressed in terms of its eigenvalues and eigenvectors.

- Applicable primarily to square matrices, and especially useful for symmetric matrices.
- If  $A$  is a symmetric matrix, it can be decomposed as:

$$A = Q\Lambda Q^T$$

where:

- $Q$  is an orthogonal matrix containing the eigenvectors of  $A$  as columns.
- $\Lambda$  is a diagonal matrix with the eigenvalues of  $A$  on the diagonal.

# Eigenvalues and Eigenvectors

**Eigenvalues** and **eigenvectors** are fundamental in understanding the properties of a matrix.

- For a matrix  $A$ , a scalar  $\lambda$  is called an **eigenvalue** if there exists a non-zero vector  $v$  (the **eigenvector**) such that:

$$Av = \lambda v$$

- The equation can be rearranged to:

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

- The eigenvalues  $\lambda$  are found by solving the characteristic equation:

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

- Once the eigenvalues are determined, the corresponding eigenvectors can be found by solving the linear system  $(A - \lambda I)v = 0$ .