Matrix Algebra

LRS

University of Guelph

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$$\mathbf{w} = \begin{pmatrix} 72\\18\\54\\37 \end{pmatrix}$$

The transpose of **w** is

$$w' = (72 \ 18 \ 54 \ 37)$$

Matrices

$$\mathbf{A} = \begin{pmatrix} 7 & 18 & -2 & 22 \\ -16 & 3 & 55 & 1 \\ 9 & -4 & 0 & 31 \end{pmatrix}$$
$$\mathbf{B} = \begin{pmatrix} x & y+1 & x+y+z \\ a-b & c\log d & e \\ \sqrt{x-y} & (m+n)/n & p \end{pmatrix}$$

and

$$\textbf{C}=\left(\begin{array}{cc}\textbf{C}_{11} & \textbf{C}_{12}\\ \textbf{C}_{21} & \textbf{C}_{22}\end{array}\right)$$

Transpose

$$\mathbf{A} = \begin{pmatrix} 7 & 18 & -2 & 22 \\ -16 & 3 & 55 & 1 \\ 9 & -4 & 0 & 31 \end{pmatrix}$$
$$\mathbf{A}' = \begin{pmatrix} 7 & -16 & 9 \\ 18 & 3 & -4 \\ -2 & 55 & 0 \\ 22 & 1 & 31 \end{pmatrix}$$

 and

Matrix Addition

Conformable for Addition Rule

Two matrices are conformable for addition if they have the same number of rows and columns.

$$\mathbf{C} = \mathbf{A} + \mathbf{B}$$
$$c_{ij} = a_{ij} + b_{ij}$$

Example

Let

$$\mathbf{A} = \begin{pmatrix} 2 & 9 \\ 7 & -3 \\ -5 & -4 \end{pmatrix} \quad \text{and} \quad \mathbf{B} = \begin{pmatrix} 1 & -8 \\ -5 & 5 \\ 6 & 7 \end{pmatrix}$$

Both have 3 rows and 2 columns, then

$$\mathbf{C} = \left(\begin{array}{rrr} 3 & 1 \\ 2 & 2 \\ 1 & 3 \end{array}\right)$$

Types of Matrices

- Rectangular, $r \neq c$
- Square, r = c $\mathbf{P} = \begin{pmatrix} 2 & 4 & -6 & 1 \\ 1 & 3 & -5 & 0 \\ 4 & 1 & 7 & -3 \\ -2 & -1 & -4 & 8 \end{pmatrix}$ • Symmetric ($\mathbf{T} = \mathbf{T}'$)

$$\mathbf{T} = \begin{pmatrix} 2 & 4 & -6 & 1 \\ 4 & 3 & -5 & 0 \\ -6 & -5 & 7 & -3 \\ 1 & 0 & -3 & 8 \end{pmatrix}$$

More Types

Diagonal

$$\mathbf{D} = \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 \\ 0 & 0 & 7 & 0 \\ 0 & 0 & 0 & 8 \end{pmatrix}$$

- Identity
- Null
- Matrices With Only 1's
- Triangular, upper or lower

$$\mathbf{T} = \begin{pmatrix} 2 & 4 & -6 & 1 \\ 0 & 3 & -5 & 0 \\ 0 & 0 & 7 & -3 \\ 0 & 0 & 0 & 8 \end{pmatrix}$$

One More Type

Tridiagonal

$$\mathbf{B} = \begin{pmatrix} 10 & 3 & 0 & 0 & 0 & 0 \\ 3 & 10 & 3 & 0 & 0 & 0 \\ 0 & 3 & 10 & 3 & 0 & 0 \\ 0 & 0 & 3 & 10 & 3 & 0 \\ 0 & 0 & 0 & 3 & 10 & 3 \\ 0 & 0 & 0 & 0 & 3 & 10 \end{pmatrix}$$

Matrix Multiplication

Conformable for Multiplication Rule

Two matrices are conformable for multiplication if the number of columns in the first matrix equals the number of rows in the second matrix.

$$\mathbf{C}_{p\times q} = \{c_{ij}\}$$

and

$$\mathsf{D}_{m\times n} = \{d_{ij}\}$$

and q = m, then

$$\mathsf{CD}_{p\times n} = \{\sum_{k=1}^m c_{ik} d_{kj}\}$$

Example

$$\mathbf{C} = \begin{pmatrix} 6 & 4 & -3 \\ 3 & 9 & -7 \\ 8 & 5 & -2 \end{pmatrix} \text{ and } \mathbf{D} = \begin{pmatrix} 1 & 1 \\ 2 & 0 \\ 3 & -1 \end{pmatrix}$$
$$\mathbf{CD} = \begin{pmatrix} 6(1) + 4(2) - 3(3) & | & 6(1) + 4(0) - 3(-1) \\ 3(1) + 9(2) - 7(3) & | & 3(1) + 9(0) - 7(-1) \\ 8(1) + 5(2) - 2(3) & | & 8(1) + 5(0) - 2(-1) \end{pmatrix} = \begin{pmatrix} 5 & 9 \\ 0 & 10 \\ 12 & 10 \end{pmatrix}$$

The transpose of the product of two or more matrices is the product of the transposes of each matrix in reverse order.

 $(CDE)'\ =\ E'D'C'$

Special Products

Idempotent

$$AA = A$$

Nilpotent

$$\mathbf{BB} = \mathbf{0}$$

Orthogonal

$$UU' = I$$

which also implies that

 $\mathbf{U}'\mathbf{U} = \mathbf{I}$

provided that **U** is square.

- Multiplication of a matrix by a scalar results in multiplying every element of the matrix by that scalar.

- $(\mathbf{A} + \mathbf{B})^2 = (\mathbf{A} + \mathbf{B})(\mathbf{A} + \mathbf{B}) = \mathbf{A}\mathbf{A} + \mathbf{A}\mathbf{B} + \mathbf{B}\mathbf{A} + \mathbf{B}\mathbf{B}.$

Traces

Traces of square matrices only.

Let

$$\mathbf{T} = \begin{pmatrix} 2 & 4 & -6 & 1 \\ 4 & 3 & -5 & 0 \\ -6 & -5 & 7 & -3 \\ 1 & 0 & -3 & 8 \end{pmatrix}$$

then

$$tr(\mathbf{T}) = 2 + 3 + 7 + 8 = 20$$

Traces are associated with degrees of freedom in hypothesis testing.



Rotation Rule

$$tr(ABC) = tr(BCA) = tr(CAB)$$

Determinants

Determinants exist only for square matrices. Let

$$\mathbf{A} = \begin{pmatrix} 6 & 2 \\ 1 & 4 \end{pmatrix}$$
$$|\mathbf{A}| = (6)(4) - (1)(2) = 22$$

General Expression

$$\mid \mathbf{A} \mid = \sum_{j=1}^{n} (-1)^{i+j} a_{ij} \mid \mathbf{M}_{ij} \mid$$

where \mathbf{M}_{ij} is the minor of element a_{ij} obtained by deleting the i^{th} row and j^{th} column of \mathbf{A} .

Matrix Algebra Review

Example Determinant, 3×3 Matrix

$$\mathbf{A} = \begin{pmatrix} 5 & 2 & 7 \\ 6 & 1 & 8 \\ 4 & 3 & 9 \end{pmatrix}$$
$$|\mathbf{A}| = 5 \begin{vmatrix} 1 & 8 \\ 3 & 9 \end{vmatrix} - 2 \begin{vmatrix} 6 & 8 \\ 4 & 9 \end{vmatrix} + 7 \begin{vmatrix} 6 & 1 \\ 4 & 3 \end{vmatrix}$$
$$|\mathbf{A}| = 5(-15) - 2(22) + 7(14) = -21$$

Any row or column of \mathbf{A} can be used, and the same value of the determinant will be obtained.

Inverse of a square matrix, A, with a non-zero determinant is denoted by

$$\mathbf{A}^{-1}$$

and satisfies

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

The inverse of **A** is calculated as

$$\mathbf{A}^{-1} = |\mathbf{A}|^{-1}\mathbf{M'}_{A}$$

where \mathbf{M}_A is the matrix of determinants of signed minors of \mathbf{A} .

Example Inversion

Let

$$\textbf{A} = \; \left(\begin{array}{ccc} 6 & -1 & 2 \\ 3 & 4 & -5 \\ 1 & 0 & -2 \end{array} \right) \;$$

Then

$$| \mathbf{A} | = -57$$

The determinants of signed minors are

$$\mathbf{M}_{\mathcal{A}} = \left(\begin{array}{rrrr} -8 & 1 & -4 \\ -2 & -14 & -1 \\ -3 & 36 & 27 \end{array}\right)$$

and

$$\mathbf{A}^{-1} = \frac{1}{-57} \begin{pmatrix} -8 & -2 & -3 \\ 1 & -14 & 36 \\ -4 & -1 & 27 \end{pmatrix}$$

Determinants of Products

Two square matrices, A and B, with the same dimensions, then

$$\mathsf{AB} \mid = \mid \mathsf{A} \mid \cdot \mid \mathsf{B} \mid$$

If $|\mathbf{A}| = 0$ or $|\mathbf{B}| = 0$, then

$$\mathbf{AB} \mid = 0$$

If $|\mathbf{AB}| \neq 0$, then

$$(\mathbf{AB})^{-1} = \mathbf{B}^{-1}\mathbf{A}^{-1}$$

Linear Dependence

Rank

Rank of any matrix is the number of linearly independent rows and columns, a scalar number.

Rules

- If the determinant of a square matrix is NOT zero, then rank is equal to the order of the matrix, full rank, and the inverse of the matrix exists.
- If the determinant of a square matrix IS zero, or if the matrix is not square, then rank is not full, and an inverse of the matrix does NOT exist.
- So If a matrix has r rows and c columns and r < c, then the rank of the matrix can not be greater than r.

Examples of Dependence

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$$\mathbf{B} = \begin{pmatrix} 3 & 9 \\ 1 & 3 \end{pmatrix}$$

Note that column 2 equals 3 times column 1, or that row 1 equals 3 times row 2. The determinant of \mathbf{B} is 0.

$$\mathbf{C} = \begin{pmatrix} 2 & -1 & 1 \\ 4 & -6 & -2 \\ -3 & 7 & 4 \end{pmatrix}$$

Note that column 3 equals the sum of column 1 and column 2 - a linear dependency. The determinant of **C** is 0. The rank of **C** can be 0, 1, or 2, but not 3, and not greater than 3.

• Not all dependencies can be spotted easily by visual observation.

Example Rank

$$\mathbf{A} = \begin{pmatrix} 1 & 3 & 5 & 2 \\ 6 & 9 & 5 & 1 \\ 8 & 7 & 0 & 4 \\ 3 & 1 & 0 & 5 \end{pmatrix}$$

which has a 0 determinant. Therefore, the rank is less than 4. Must use elementary operators to reduce matrix to a triangular matrix using **Elementary Operator** matrices. The number of non-zero diagonal elements in the reduced matrix gives the rank of the matrix. Elementary operator matrices are identity matrices that have been modified by one of three methods.

- Let E₁₁(.25) be an elementary operator where the first diagonal element of an identity matrix has been changed to .25.
- Let E_{ij} be an elementary operator where rows i and j are interchanged.
- Let E_{ij}(c) be an elementary operator where c occupies row i and column j.

Multiplying a matrix by an elementary operator matrix does not change the RANK of the matrix.

Calculation of Rank, Example

$$\mathbf{A} = \begin{pmatrix} 1 & 3 & 5 & 2 \\ 6 & 9 & 5 & 1 \\ 8 & 7 & 0 & 4 \\ 3 & 1 & 0 & 5 \end{pmatrix}$$

The goal is to reduce A to an upper triangular form through successive pre-multiplications by elementary operator matrices. The first one is

$$\mathbf{P}_1 = \begin{pmatrix} 1 & 0 & 0 & 0 \\ -6 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}, \quad \mathbf{P}_1 \mathbf{A} = \begin{pmatrix} 1 & 3 & 5 & 2 \\ 0 & -9 & -25 & -11 \\ 8 & 7 & 0 & 4 \\ 3 & 1 & 0 & 5 \end{pmatrix}$$

The same type of elementary operators can be used to eliminate the 8 in row 3 and the 3 in row 4.

Calculation of Rank

The reduced form after premultiplication by 3 elementary operators is

$$\mathbf{P}_3 \mathbf{P}_2 \mathbf{P}_1 \mathbf{A} = \begin{pmatrix} 1 & 3 & 5 & 2 \\ 0 & -9 & -25 & -11 \\ 0 & -17 & -40 & -12 \\ 0 & -8 & -15 & -1 \end{pmatrix}$$

Now two more elementary operators will make -17 in row 3 and -8 in row 4 change to 0. Finally the sixth elementary operator will change the element in the 4^{th} row and 3^{rd} column equal to 0. The final reduced matrix is

$$\mathbf{P}_{6}\mathbf{P}_{5}\mathbf{P}_{4}\mathbf{P}_{3}\mathbf{P}_{2}\mathbf{P}_{1}\mathbf{A} = \begin{pmatrix} 1 & 3 & 5 & 2 \\ 0 & -9 & -25 & -11 \\ 0 & 0 & \frac{65}{9} & \frac{79}{9} \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

The number of non-zero diagonal elements in this matrix is 3. Therefore, the rank of **A** is 3, or $r(\mathbf{A}) = 3$.

More About Rank

a
$$A_{10\times 50}$$
, then $r(A) \le 10$
a $r(AA') = r(A)$
a $A_{6\times 3}B_{3\times 10} = C_{6\times 10}$, then $r(C) \le 3$
a $r(J) = 1$
b $r(0) = 0$

- Full-row rank If **A** has order $m \times n$ with rank equal to m, then **A** has full row rank.
- Full-column rank A matrix with rank equal to the number of columns has full-column rank.
 - Full rank A square matrix with rank equal to the number of rows or columns has full rank. A full rank matrix is nonsingular, has a non-zero determinant, and has an inverse.

Partitioning A Matrix

A of order $p \times q$, rank r, and r is less than or equal to the smaller of p or q.

$$\mathbf{A}_{p\times q} = \left(\begin{array}{cc} \mathbf{A}_{11} & \mathbf{A}_{12} \\ \mathbf{A}_{21} & \mathbf{A}_{22} \end{array}\right)$$

A₁₁ order $r \times r$ and rank r. Re-arrangement of rows and columns of **A** may be needed to find an appropriate **A**₁₁. **A**₁₂ order $r \times (q - r)$ **A**₂₁ order $(p - r) \times r$, and **A**₂₂ order $(p - r) \times (q - r)$

Dependencies in Partitions

$$\begin{array}{l} \mathsf{A}_{21} \quad \mathsf{A}_{22} \end{array} \big) = \mathsf{K}_2 \left(\begin{array}{cc} \mathsf{A}_{11} \quad \mathsf{A}_{12} \end{array} \right) \\ \\ \left(\begin{array}{c} \mathsf{A}_{12} \\ \mathsf{A}_{22} \end{array} \right) = \left(\begin{array}{c} \mathsf{A}_{11} \\ \mathsf{A}_{21} \end{array} \right) \mathsf{K}_1 \\ \\ \mathsf{A} = \left(\begin{array}{c} \mathsf{A}_{11} \quad \mathsf{A}_{11}\mathsf{K}_1 \\ \mathsf{K}_2\mathsf{A}_{11} \quad \mathsf{K}_2\mathsf{A}_{11}\mathsf{K}_1 \end{array} \right) \end{array}$$

Generalized Inverses

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

Moore-Penrose inverse satisfies:

- $\bullet AA^{-}A = A,$
- $(\mathbf{A}^{-}\mathbf{A})' = \mathbf{A}^{-}\mathbf{A}, \text{ and }$

Compute the generalized inverse as

$$\mathbf{A}^{-}=\left(\begin{array}{cc}\mathbf{A}_{11}^{-1} & \mathbf{0}\\ \mathbf{0} & \mathbf{0}\end{array}\right)$$

If the generalized inverse satisfies only one of the Moore-Penrose conditions, then there are an infinite number of generalized inverses for any non full rank matrix.

If ${\bf G}$ represents the generalized inverse, then other generalized inverses, ${\bf F},$ can be obtained from

$$\mathbf{F} = \mathbf{G}\mathbf{A}\mathbf{G} + (\mathbf{I} - \mathbf{G}\mathbf{A})\mathbf{X} + \mathbf{Y}(\mathbf{I} - \mathbf{A}\mathbf{G})$$

for any arbitrary matrices **X** and **Y** of the correct order.

Other Things in Notes

- Eigenvalues and eigenvectors
- Differentiation, tomorrow
- Cholesky decomposition