

4.1. Matrix Operations

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1j} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2j} & \dots & a_{2n} \\ \vdots & \vdots & & & & \vdots \\ a_{i1} & a_{i2} & \dots & a_{ij} & \dots & a_{in} \\ \vdots & \vdots & & & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mj} & \dots & a_{mn} \end{bmatrix}$$

• The entry in the i th row and the j th column of a matrix A is referred to as $(A)_{ij}$.

EXAMPLE:

• A **zero matrix** is a matrix, written 0 , whose entries are all zero.

• A **square** matrix has the same number of rows than columns.

• In general ($m \neq n$), matrices are **rectangular**.

• The **(main) diagonal** of a matrix, or its **diagonal entries**, are the entries

• A **diagonal matrix** has all its nondiagonal entries equal to zero.

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

• A matrix is **upper triangular** if all its elements under the diagonal are zero

• A matrix is **lower triangular** if all its elements over the diagonal are zero

• The set of all possible matrices of dimension $(m \times n)$ whose entries are real numbers is referred to as $\mathbb{R}^{m \times n}$

• The set of all possible matrices of dimension $(m \times n)$ whose entries are complex numbers is referred to as $\mathbb{C}^{m \times n}$

$$\begin{bmatrix} 0 & 1 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & -1 \end{bmatrix} \quad \begin{bmatrix} 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 4 \end{bmatrix} \quad \begin{bmatrix} 2 & 2 \\ 7 & 1 \\ 3 & -3 \end{bmatrix} \in \mathbb{K}^{3 \times 2}$$

• OPERATIONS:

Only for matrices with the same dimensions:

o **Equality.** Two matrices are equal if and only if their corresponding entries are equal.

$$\begin{bmatrix} 3 & -1 \\ 1 & 0 \end{bmatrix} \neq \begin{bmatrix} & \\ & \end{bmatrix} \neq \begin{bmatrix} & \\ & \end{bmatrix}$$

o **Addition.** A matrix whose entries are the sum of the corresponding entries of the matrices.

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

- **Scalar Multiplication.** A matrix whose entries are the corresponding entries of the matrix multiplied by the scalar.

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

● **PROPERTIES:**

Let A, B and C be matrices of $\mathbb{K}^{m \times n}$ and $\lambda, \mu \in \mathbb{K}$:

- $A + B = B + A$ ○ $\lambda(A+B) = \lambda A + \lambda B$
- $A+(B+C)=(A+B)+C$ ○ $(\lambda+\mu)A = \lambda A + \mu A$
- $A + 0 = A$ ○ $\lambda(\mu A) = (\lambda\mu)A$

Matrix Multiplication

$$\boxed{\mathbb{K}^p}$$

$$\boxed{\mathbb{K}^n}$$

$$\boxed{\mathbb{K}^m}$$

One wonders:

$$\text{Does } C \text{ exist} \mid C\mathbf{x} = A B \mathbf{x} \quad \forall \mathbf{x} \in \mathbb{K}^p ?$$

PROBLEM: *What dimensions would C have?*

If we write $B = [\mathbf{b}_1 \mathbf{b}_2 \dots \mathbf{b}_p]$ and $\mathbf{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_p \end{bmatrix}$, then:

$$B\mathbf{x} = x_1 \mathbf{b}_1 + x_2 \mathbf{b}_2 + \dots + x_p \mathbf{b}_p$$

$$A(B\mathbf{x}) =$$

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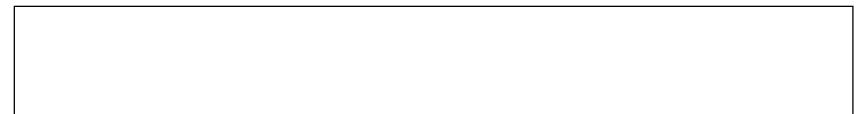
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● Let A be an $(m \times n)$ matrix and let B be an $(n \times p)$ matrix with columns $\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_p$. The **matrix product** of A by B is the $(m \times p)$ matrix AB whose columns are $A\mathbf{b}_1, A\mathbf{b}_2, \dots, A\mathbf{b}_p$.

That is,

$$AB = A [\mathbf{b}_1 \mathbf{b}_2 \dots \mathbf{b}_p] = [A\mathbf{b}_1 A\mathbf{b}_2 \dots A\mathbf{b}_p]$$



Warning: The dimensions of the matrices involved in a product must verify

$$\boxed{\begin{matrix} A & B & = & C \\ (&) & (&) & (&) \end{matrix}}$$

EXAMPLE:

$$\begin{aligned} \begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} 4 & 3 & 6 \\ 1 & -2 & 3 \end{bmatrix} &= \\ \left(\quad \right) \quad \left(\quad \right) &\Rightarrow \left(\quad \right) \\ \\ = \left[\begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix} \left[\quad \right] \quad \begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix} \left[\quad \right] \quad \begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix} \left[\quad \right] \right] &= \\ = \left[\left[\quad \right] \quad \left[\quad \right] \quad \left[\quad \right] \right] &= \left[\quad \right] \end{aligned}$$

EXAMPLE:

$$\begin{aligned} \begin{bmatrix} \boxed{2} & \boxed{3} \\ 1 & -5 \end{bmatrix} \begin{bmatrix} 4 & 3 & \boxed{6} \\ 1 & -2 & \boxed{3} \end{bmatrix} &= \begin{bmatrix} \star & \star & \star \\ \star & \star & \star \end{bmatrix} \\ \text{1st row} \quad \quad \text{3rd column} &\rightarrow \text{(1, 3) entry} \\ \\ \begin{bmatrix} \boxed{2} & \boxed{3} \\ \boxed{1} & -5 \end{bmatrix} \begin{bmatrix} \boxed{4} & 3 & 6 \\ 1 & -2 & 3 \end{bmatrix} &= \begin{bmatrix} \star & \star & \star \\ \star & \star & \star \end{bmatrix} \\ \text{2nd row} \quad \quad \text{1st column} &\rightarrow \text{(2, 1) entry} \end{aligned}$$

Row-Column Rule for computing AB

Consider $A \in \mathbb{K}^{m \times n}$, and $B = [\mathbf{b}_1 \dots \mathbf{b}_p] \in \mathbb{K}^{n \times p}$ such that $(A)_{ik} = a_{ik}$, and $(B)_{kj} = b_{kj}$.

$$AB = \underbrace{[\mathbf{A}\mathbf{b}_1 \quad \dots \quad \mathbf{A}\mathbf{b}_j \quad \dots \quad \mathbf{A}\mathbf{b}_p]}_{\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & \vdots & & \vdots \\ a_{i1} & a_{i2} & \dots & a_{in} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} b_{1j} \\ b_{2j} \\ \vdots \\ b_{nj} \end{bmatrix} = \begin{bmatrix} \star_1 \\ \vdots \\ \star_i \\ \vdots \\ \star_m \end{bmatrix} \rightarrow (AB)_{ij}$$

That is,

$$(AB)_{ij} = \begin{bmatrix} a_{i1} & a_{i2} & \dots & a_{in} \end{bmatrix} \begin{bmatrix} b_{1j} \\ b_{2j} \\ \vdots \\ b_{nj} \end{bmatrix} = \sum_k a_{ik} b_{kj}$$

PROBLEM: Find the 2nd row of AB .

$$AB = \begin{bmatrix} 2 & -5 & 0 \\ -1 & 3 & -4 \\ 6 & -8 & -7 \\ -3 & 0 & 9 \end{bmatrix} \begin{bmatrix} 4 & -6 \\ 7 & 1 \\ 3 & 2 \end{bmatrix}$$

PROBLEM: Compute

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

• **PROPERTIES:**

Let A be an $(m \times n)$ matrix, and B and C matrices of appropriate dimensions:

- $A(BC) = (AB)C$
- $A(B + C) = AB + AC$
- $(B + C)A = BA + CA$
- $\mu(AB) = (\mu A)B = A(\mu B) \quad \forall \mu \in \mathbb{K}$
- $\mathbb{I}_m A = A = A \mathbb{I}_n$ where \mathbb{I}_k is the $(k \times k)$ **identity matrix**

→ 4.3

WARNING: In general, $AB = AC \not\Rightarrow B = C$

ROTATION $\pi/2$

$$B = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$

PROJECTION in X

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

1st ROTATION + 2nd PROJECTION

$$AB = \begin{bmatrix} & \\ & \end{bmatrix}$$

REFLECTION $x+y=0$

$$C = \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix}$$

PROJECTION in X

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

1st REFLECTION + 2nd PROJECTION

$$AC = \begin{bmatrix} & \\ & \end{bmatrix}$$

WARNING: In general, $AB \neq BA$

EXPANSION AXIS X

$$B = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$$

ROTATION 30°

$$A = \frac{1}{2} \begin{bmatrix} \sqrt{3} & -1 \\ 1 & \sqrt{3} \end{bmatrix}$$

1st EXPANSION + 2nd ROTATION

$$AB = \begin{bmatrix} & \\ & \end{bmatrix}$$

ROTATION 30°

$$A = \frac{1}{2} \begin{bmatrix} \sqrt{3} & -1 \\ 1 & \sqrt{3} \end{bmatrix}$$

EXPANSION AXIS X

$$B = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$$

1st ROTATION + 2nd EXPANSION

$$BA = \begin{bmatrix} & \\ & \end{bmatrix}$$

WARNING: In general, $AB = 0 \not\Rightarrow A = 0$ or $B = 0$

PROJECTION in X

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

PROJECTION in Y

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

1st X-PROJECTION + 2nd Y-PROJECTION

$$AB = \begin{bmatrix} & \\ & \end{bmatrix}$$

WARNING: In general, $A^2 = 0 \not\Rightarrow A = 0$

$$A = \begin{bmatrix} 1 & 1 \\ -1 & -1 \end{bmatrix}, \quad A = \begin{bmatrix} 1 & 1 \\ -1 & -1 \end{bmatrix} \Rightarrow A^2 = \begin{bmatrix} & \\ & \end{bmatrix}$$

• If two square matrices verify that $AB = BA$, we say that A and B **commute** with each other.

- The k th **power** of a matrix is defined:

$$A^k = \underbrace{A A A \cdots A}_{k \text{ times}}$$

This only makes sense if A is a _____ matrix and k is a nonnegative integer.

- For convenience, we define $A^0 = \mathbb{I}$.

PROBLEM: Compute

→ 4.4

Transpose of a Matrix

- The **transpose** of an $(m \times n)$ matrix A is the $(n \times m)$ matrix A^T whose columns are the rows of A .

That is,

$$(A^T)_{ij} = (A)_{ji}$$

EXAMPLE:

$$B = \begin{bmatrix} -5 & 1 & 0 \\ 2 & -3 & 4 \end{bmatrix} \Rightarrow B^T =$$

EXAMPLE:

- A **symmetric** matrix verifies $A^T = A$.
- An **antisymmetric** matrix verifies $A^T = -A$.

PROBLEM: Provide examples of (anti)symmetric matrices.

• PROPERTIES:

Let A and B be matrices of appropriate dimensions and $\mu \in \mathbb{K}$:

- $(A^T)^T = A$
- $(A + B)^T = A^T + B^T$
- $(\mu A)^T = \mu (A^T)$
- $(AB)^T = B^T A^T$

Proof: Let be $A \in \mathbb{K}^{m \times n}$ and $B \in \mathbb{K}^{n \times q}$

$$((AB)^T)_{ij} =$$

PROBLEM: Prove that $(ABC)^T = C^T B^T A^T$.

→ 4.7

Conjugate Transpose of a Matrix

- The **conjugate transpose** of an $(m \times n)$ matrix A is the $(n \times m)$ matrix A^* , or A^H , whose elements verify:

$$(A^*)_{ij} = \overline{(A)_{ji}}$$

EXAMPLE:

$$B = \begin{bmatrix} -5 & 2-i \\ i & 3 \\ 0 & 4 \end{bmatrix} \Rightarrow B^* =$$

$$A = [\mathbf{a}_1 \ \mathbf{a}_2 \ \cdots \ \mathbf{a}_n] \Rightarrow A^* =$$

• **PROPERTIES:**

Let A and B be matrices of appropriate dimensions and $\mu \in \mathbb{K}$:

- $(A^*)^* = A$
- $(A + B)^* = A^* + B^*$
- $(\mu A)^* = \bar{\mu}(A^*)$
- $(AB)^* = B^* A^*$
- $A^* = A^T$ if and only if A is a real matrix.

- A **Hermitian** matrix verifies $A^* = A$.
- An **antihermitian** matrix verifies $A^* = -A$.

PROBLEM: Provide examples of (anti)Hermitian matrices.

→ 4.8

4.2. Inverse of a Matrix

• A square ($n \times n$) matrix A is **invertible**, or **nonsingular**, if there exists a matrix B such that

$$AB = \mathbb{I}_n$$

- A **noninvertible** or **singular** matrix has no inverse.

EXAMPLE: This matrix is invertible: $A = \begin{bmatrix} 2 & 5 \\ -3 & -7 \end{bmatrix}$

Because $C = \begin{bmatrix} -7 & -5 \\ 3 & 2 \end{bmatrix}$ verifies $AC = \begin{bmatrix} 2 & 5 \\ -3 & -7 \end{bmatrix} \begin{bmatrix} -7 & -5 \\ 3 & 2 \end{bmatrix} = \begin{bmatrix} & \\ & \end{bmatrix}$

EXAMPLE: This matrix is invertible: $A = \begin{bmatrix} \cos \phi & -\sin \phi \\ \sin \phi & \cos \phi \end{bmatrix}$

$$A = \boxed{\phantom{\begin{bmatrix} & \\ & \end{bmatrix}}} \Rightarrow A^{-1} = \boxed{\phantom{\begin{bmatrix} & \\ & \end{bmatrix}}}$$

Thus, $A^{-1} = \begin{bmatrix} & \\ & \end{bmatrix} = \begin{bmatrix} & \\ & \end{bmatrix}$

EXAMPLE: Matrix B has no inverse and is, therefore, a singular matrix:

$$B = \boxed{\phantom{\begin{bmatrix} & \\ & \end{bmatrix}}} = \begin{bmatrix} & \\ & \end{bmatrix}$$

→ 4.9

Theorem 4.1. If A is an invertible ($n \times n$) matrix, then the equation $A\mathbf{x} = \mathbf{b}$ has the unique solution $\mathbf{x} = A^{-1}\mathbf{b}$, $\forall \mathbf{b} \in \mathbb{K}^n$.

Proof:

- That $\mathbf{x} = A^{-1}\mathbf{b}$ is a solution $\forall \mathbf{b}$ can be checked by a mere substitution:
- As it has a solution $\forall \mathbf{b} \Rightarrow A$ must have a pivot in every row.

A square
 \Rightarrow

No free variables
 \Rightarrow

Warning:

Theorem 4.2. Let A and B be $(n \times n)$ matrices. Then:

$$AB = \mathbb{I} \iff BA = \mathbb{I}$$

Proof: $(AB = \mathbb{I} \Rightarrow BA = \mathbb{I})$

o Suppose that $BA = X$

o Let's define $M = \mathbb{I} - X = [\mathbf{m}_1 \ \mathbf{m}_2 \ \cdots \ \mathbf{m}_n]$.

As

That is,

o But now,

Leading to

Theorem 4.3. If A is an invertible matrix, then A^{-1} is invertible and $(A^{-1})^{-1} = A$.

Proof:

Theorem 4.4. If exists, the inverse of a matrix is unique.

Proof: Let A be an invertible matrix, and B a matrix such that $AB = \mathbb{I}$ (that is, $B = A^{-1}$). Suppose there exists C such that $AC = \mathbb{I}$ (in other words, suppose that A has another inverse).

Theorem 4.5. If A is invertible, A^T is also invertible and $(A^T)^{-1} = (A^{-1})^T$.

Theorem 4.6. If A is invertible, A^* is also invertible and $(A^*)^{-1} = (A^{-1})^*$.

Proof:

EXAMPLE:

$$\begin{bmatrix} 1+i & 1+2i \\ -1 & -1-i \end{bmatrix} \begin{bmatrix} -1-i & -1-2i \\ 1 & 1+i \end{bmatrix} = \begin{bmatrix} & \\ & \end{bmatrix}$$

then, $\begin{bmatrix} -1+i & 1 \\ -1+2i & 1-i \end{bmatrix}^{-1} = \begin{bmatrix} & \\ & \end{bmatrix}$

Theorem 4.7. If A and B are invertible $(n \times n)$ matrices, then AB is invertible and $(AB)^{-1} = B^{-1}A^{-1}$.

Proof:

EXAMPLE: Consider the linear transformations:

$$A = \boxed{\text{ROTATE}} \quad B = \boxed{\text{EXPAND}}.$$

Then,

$$AB = \boxed{} \boxed{} = \boxed{}$$

(in this order!) and the inverse is

$$(AB)^{-1} = \boxed{} = \boxed{}^{-1} \boxed{}^{-1}$$

PROBLEM: If A , B and C are nonsingular matrices of equal size, show that $(ABC)^{-1} = C^{-1}B^{-1}A^{-1}$.

→ 4.11

• An **elementary matrix** is one that is obtained by performing one elementary row operation on an identity matrix.

EXAMPLE:

$$E_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 5 \end{bmatrix} \quad E_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \quad E_3 = \begin{bmatrix} 1 & 0 & 0 \\ -4 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Notice: These matrices have a clear geometrical interpretation. They correspond to

Theorem 4.8. If an elementary row operation is performed on an $(m \times n)$ matrix A , the resulting matrix can be written as EA , where E is the $(m \times m)$ elementary matrix created by performing the same operation on \mathbb{I}_m .

EXAMPLE: Consider the (3×2) matrix $A = \begin{bmatrix} a & d \\ b & e \\ c & f \end{bmatrix}$

◦ $\mathbb{I} \sim E_1$ ($r_3 \rightarrow 5r_3$)

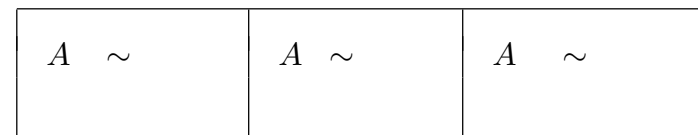
$$E_1 A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 5 \end{bmatrix} \begin{bmatrix} a & d \\ b & e \\ c & f \end{bmatrix} = \begin{bmatrix} a & d \\ b & e \\ 5c & 5f \end{bmatrix}$$

◦ $\mathbb{I} \sim E_2$ ($r_2 \leftrightarrow r_3$)

$$E_2 A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} a & d \\ b & e \\ c & f \end{bmatrix} = \begin{bmatrix} a & d \\ c & f \\ b & e \end{bmatrix}$$

◦ $\mathbb{I} \sim E_3$ ($r_2 \rightarrow r_2 - 4r_1$)

$$E_3 A = \begin{bmatrix} 1 & 0 & 0 \\ -4 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a & d \\ b & e \\ c & f \end{bmatrix} = \begin{bmatrix} a & d \\ b - 4a & e - 4d \\ c & f \end{bmatrix}$$



Theorem 4.9. Every elementary matrix E is invertible and its inverse E^{-1} is the elementary matrix corresponding to the row operation that transforms E back into \mathbb{I} .

EXAMPLE: The matrix E_1 multiplies the 3rd row by five:

$$E_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 5 \end{bmatrix}$$

Its inverse E_1^{-1} is the matrix that divides the 3rd row by five:

$$E_1^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1/5 \end{bmatrix}$$

Check: $E_1 E_1^{-1} = \dots = \mathbb{I}$

→ 4.12

PROBLEM: Find the matrices E_2^{-1} and E_3^{-1} .

Theorem 4.10. An $(n \times n)$ matrix A is invertible if and only if A is row equivalent to \mathbb{I}_n . In this case, any sequence of elementary row operations that transforms A into \mathbb{I}_n also transforms \mathbb{I}_n in A^{-1} .

Proof:

A invertible \Leftrightarrow

\Rightarrow

Then, $A^{-1} = E_p E_{p-1} \dots E_2 E_1$ and, in fact,

$\rightarrow 4.14$

An Algorithm for finding A^{-1}

- o Construct the matrix $[A \ \mathbb{I}]$
- o Find its reduced echelon form.
- o If this matrix has the form $[\mathbb{I} \ B]$, then $A^{-1} = B$. Otherwise, A does not have an inverse.

EXAMPLE:

$$\underbrace{\begin{bmatrix} 0 & 1 & 2 & 1 & 0 & 0 \\ 1 & 0 & 3 & 0 & 1 & 0 \\ 4 & -3 & 8 & 0 & 0 & 1 \end{bmatrix}}_A \sim \underbrace{\begin{bmatrix} 1 & 0 & 3 & 0 & 1 & 0 \\ 0 & 1 & 2 & 1 & 0 & 0 \\ 4 & -3 & 8 & 0 & 0 & 1 \end{bmatrix}}_{\mathbb{I}} \sim$$

$$\underbrace{\begin{bmatrix} 1 & 0 & 3 & 0 & 1 & 0 \\ 0 & 1 & 2 & 1 & 0 & 0 \\ 0 & -3 & -4 & 0 & -4 & 1 \end{bmatrix}}_{\mathbb{I}} \sim \underbrace{\begin{bmatrix} 1 & 0 & 3 & 0 & 1 & 0 \\ 0 & 1 & 2 & 1 & 0 & 0 \\ 0 & 0 & 2 & 3 & -4 & 1 \end{bmatrix}}_{\mathbb{I}} \sim$$

$$\begin{bmatrix} 1 & 0 & 3 & 0 & 1 & 0 \\ 0 & 1 & 0 & -2 & 4 & -1 \\ 0 & 0 & 2 & 3 & -4 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 3 & 0 & 1 & 0 \\ 0 & 1 & 0 & -2 & 4 & -1 \\ 0 & 0 & 1 & \frac{3}{2} & -2 & \frac{1}{2} \end{bmatrix} \sim$$

$$\underbrace{\begin{bmatrix} 1 & 0 & 0 & -\frac{9}{2} & 7 & -\frac{3}{2} \\ 0 & 1 & 0 & -2 & 4 & -1 \\ 0 & 0 & 1 & \frac{3}{2} & -2 & \frac{1}{2} \end{bmatrix}}_{\mathbb{I}} \Rightarrow A^{-1} = \begin{bmatrix} & & & & & \\ & & & & & \\ & & & & & \end{bmatrix}$$

PROBLEM: If exists, find the inverse of the matrix

$$C = \begin{bmatrix} 1 & 0 & -2 \\ 3 & 1 & -2 \\ -5 & -1 & 9 \end{bmatrix}$$

$$[C \ \mathbb{I}] = \begin{bmatrix} & & & & & \\ & & & & & \\ & & & & & \end{bmatrix}$$

Check: $CC^{-1} =$

$\rightarrow 4.16$

Theorem 4.11. (The Square Matrix Theorem)

If $A \in \mathbb{K}^{n \times n}$, the following statements are equivalent:

1. A is an invertible matrix.
2. There exists $C \in \mathbb{K}^{n \times n}$ such that $AC = \mathbb{I}_n$.
3. There exists $D \in \mathbb{K}^{n \times n}$ such that $DA = \mathbb{I}_n$.
4. A is row equivalent to \mathbb{I}_n .
5. A has n pivots.
6. The equation $Ax = \mathbf{0}$ has only the trivial solution.
7. The columns/rows of A are linearly independent.
8. The equation $Ax = \mathbf{b}$ has a (unique) solution $\forall \mathbf{b} \in \mathbb{K}^n$.
9. The columns/rows of A span \mathbb{K}^n .
10. The columns/rows of A form a basis of \mathbb{K}^n .
11. A^T is invertible.
12. A^* is invertible.
13. The linear transformation $\mathbf{x} \rightarrow Ax$ is bijective.
14. $\text{Col } A = \text{Row } A = \mathbb{K}^n$
15. $\dim \text{Col } A = \dim \text{Row } A = n$
16. $\text{rank } A = n$
17. $\text{Nul } A = \{\mathbf{0}\}$
18. $\dim \text{Nul } A = 0$

4.3. Partitioned (or Block) Matrices

EXAMPLE:

$$A = \left[\begin{array}{ccc|cc|c} 3 & 0 & -1 & 5 & 9 & -2 \\ -5 & 2 & 4 & 0 & -3 & 1 \\ -8 & -6 & 3 & 1 & 7 & -4 \end{array} \right]$$

$$A = \left[\begin{array}{|c|c|c|} \hline & & \\ \hline & & \\ \hline & & \\ \hline \end{array} \right] = \left[\begin{array}{|c|c|c|} \hline & & \\ \hline & & \\ \hline & & \\ \hline \end{array} \right]$$

where

$$A_{11} = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}, \quad A_{12} = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}, \quad A_{13} = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

$$A_{21} = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}, \quad A_{22} = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}, \quad A_{23} = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

• A transformation $T : \mathbb{K}^n \rightarrow \mathbb{K}^n$ is called **invertible** if there exists a transformation $S : \mathbb{K}^n \rightarrow \mathbb{K}^n$ such that

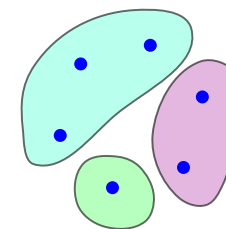
$$\left. \begin{array}{l} S(T(\mathbf{x})) = \mathbf{x} \\ T(S(\mathbf{x})) = \mathbf{x} \end{array} \right\} \quad \forall \mathbf{x} \in \mathbb{K}^n.$$

The transformation S is called the **inverse** of T .

Theorem 4.12. Let $T : \mathbb{K}^n \rightarrow \mathbb{K}^n$ be a linear transformation and A its canonical matrix. T is invertible if and only if A is nonsingular. In this case, $S(\mathbf{x}) = A^{-1}\mathbf{x}$.

→ 4.17

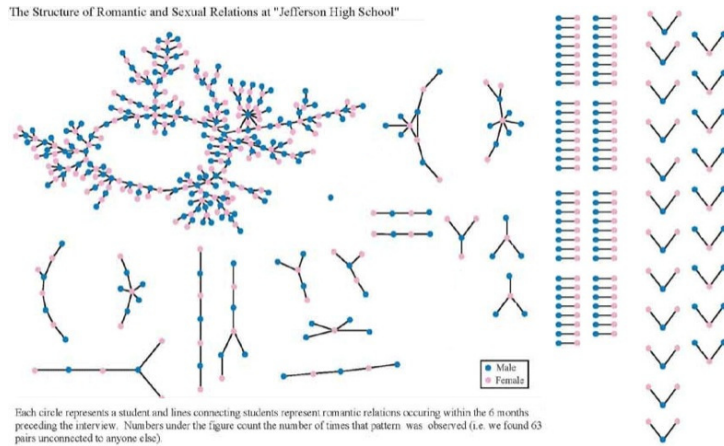
EXAMPLE: Social web of 6 persons in 3 groups



Adjacency Matrix

$$M = \left[\begin{array}{ccc|ccc} 0 & 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 & 0 \end{array} \right] = \begin{bmatrix} M_{11} & M_{12} & M_{13} \\ M_{21} & M_{22} & M_{23} \\ M_{31} & M_{32} & M_{33} \end{bmatrix}$$

EXAMPLE: *Jefferson High School*



• PROPERTIES:

- o **Addition:** Matrices of equal size and identical partition can be summed block by block:

$$A + B = \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \end{bmatrix} + \begin{bmatrix} B_{11} & B_{12} & B_{13} \\ B_{21} & B_{22} & B_{23} \end{bmatrix} = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

- o **Scalar Multiplication:**

$$\lambda A = \lambda \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \end{bmatrix} = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

EXAMPLE: *Trade share matrix between countries*

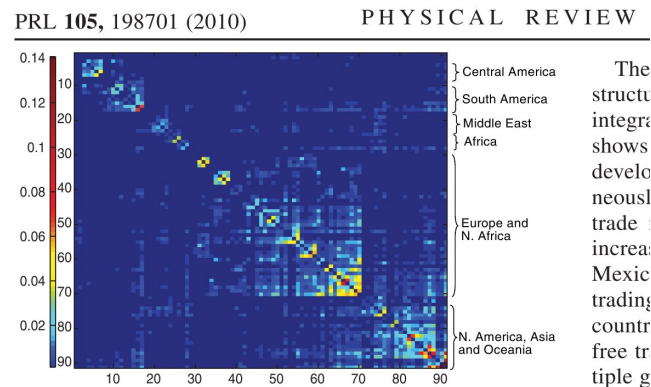


FIG. 3 (color online). The trade share matrix $S_{ij} = M_{ij} / (\sum_{m=1}^N M_{im} + \sum_{n=1}^N M_{jn})$ after hierarchical clustering between countries in 2007. We can see clearly several modules:

- o **Transpose of a matrix:**

$$A = \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \end{bmatrix} \Rightarrow A^T = \begin{bmatrix} A_{11}^T & A_{21}^T \\ A_{12}^T & A_{22}^T \\ A_{13}^T & A_{23}^T \end{bmatrix} \neq \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

- o **Conjugate transpose of a matrix:**

$$A = \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \end{bmatrix} \Rightarrow A^* = \begin{bmatrix} A_{11}^* & A_{21}^* \\ A_{12}^* & A_{22}^* \\ A_{13}^* & A_{23}^* \end{bmatrix}$$

EXAMPLE:

$$A = \left[\begin{array}{cc|c} 2 & 0 & 8 \\ 1 & -5 & 3 \\ \hline 0 & -2 & 7 \end{array} \right] \Rightarrow A^T = \left[\begin{array}{cc|c} 2 & 1 & 0 \\ 0 & -5 & -2 \\ \hline 8 & 3 & 7 \end{array} \right]$$

- o **Multiplication of partitioned matrices:** Two matrices A and B of respective dimensions $(m \times n)$ and $(n \times p)$ are conformable for block multiplication when the number of columns of each partition of A is equal to the number of rows of the corresponding partition of B .

$$AB = \left[\begin{array}{ccc|cc} 2 & -3 & 1 & 0 & -4 \\ 1 & 5 & -2 & 3 & -1 \\ \hline 0 & -4 & -2 & 7 & -1 \end{array} \right] \left[\begin{array}{c} 6 & 4 \\ -2 & 1 \\ -3 & 7 \\ -1 & 3 \\ 5 & 2 \end{array} \right]$$

$$= \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} B_{11} \\ B_{21} \end{bmatrix} = \begin{bmatrix} \phantom{A_{11}} \\ \phantom{A_{21}} \end{bmatrix}$$

(Attention:

EXAMPLE: Let A be a block upper triangular matrix:

$$A = \begin{bmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{bmatrix}.$$

Assuming that A is invertible, A_{11} is $(p \times p)$ and A_{22} is $(q \times q)$, find a formula for A^{-1} .

Call $B = A^{-1}$. Partition B in such a way that we can write:

$$AB = \begin{bmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{bmatrix} \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} = \begin{bmatrix} \mathbb{I} & 0 \\ 0 & \mathbb{I} \end{bmatrix}.$$

The dimensions of the matrices involved are:

$$\begin{bmatrix} (p \times p) & (\quad) \\ (\quad) & (q \times q) \end{bmatrix} \begin{bmatrix} (\quad) & (\quad) \\ (\quad) & (\quad) \end{bmatrix} = \begin{bmatrix} (\quad) & (\quad) \\ (\quad) & (\quad) \end{bmatrix}.$$

Concentrate on the dimensions of the blocks:

$$[(3 \times 5)] [(5 \times 2)] = \begin{bmatrix} (2 \times 3) & (\quad) \\ (\quad) & (\quad) \end{bmatrix} \begin{bmatrix} (\quad) \\ (\quad) \end{bmatrix} =$$

$$= \begin{bmatrix} (2 \times 3)(3 \times 2) + (\quad)(\quad) \\ (\quad)(\quad) + (\quad)(\quad) \end{bmatrix} =$$

$$= \begin{bmatrix} (2 \times 2) + (\quad) \\ (\quad) + (\quad) \end{bmatrix} = \begin{bmatrix} (\quad) \\ (\quad) \end{bmatrix} = [(\quad)]$$

The equation can be written:

$$\begin{bmatrix} \phantom{A_{11}} & \phantom{A_{12}} \\ \phantom{A_{21}} & \phantom{A_{22}} \end{bmatrix} = \begin{bmatrix} \mathbb{I} & 0 \\ 0 & \mathbb{I} \end{bmatrix}.$$

Equating components, we obtain:

$$\begin{aligned} \text{(a)} &= \mathbb{I} \\ \text{(b)} &= 0 \\ \text{(c)} &= 0 \\ \text{(d)} &= \mathbb{I} \end{aligned}$$

We have to solve 4 matrix equations, which represent a linear system of $(p + q)^2$ equations with $(p + q)^2$ unknowns.

o (d)

o (c)

o (a)

o (b)

Obtaining,

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

Theorem 4.14. A diagonal matrix is invertible if and only if none of its diagonal elements is zero.

$$\begin{bmatrix} a_{11} & 0 & \dots & 0 \\ 0 & a_{22} & & 0 \\ \vdots & & \ddots & \\ 0 & 0 & & a_{nn} \end{bmatrix}^{-1} = \begin{bmatrix} & & & \\ & & & \\ & & & \\ & & & \end{bmatrix}$$

PROBLEM: Determine under what conditions the following matrix is invertible and, in that case, find its inverse:

$$\begin{bmatrix} \mathbb{I}_m & 0 \\ A & \mathbb{I}_n \end{bmatrix}.$$

→ 4.19

Theorem 4.13. A block diagonal matrix is invertible if and only if each of the diagonal blocks is invertible.

Proof: The case of two blocks follows from the above result when $A_{12} = 0$.

$$\begin{bmatrix} C_{11} & 0 & \dots & 0 \\ 0 & C_{22} & & 0 \\ \vdots & & \ddots & \\ 0 & 0 & & C_{nn} \end{bmatrix}^{-1} = \begin{bmatrix} & 0 & \dots & 0 \\ 0 & & & 0 \\ \vdots & & \ddots & \\ 0 & 0 & & \end{bmatrix}$$

4.4. Determinants

• Given an $(m \times n)$ matrix A , we define the **minor** A_{ij} as the $((m-1) \times (n-1))$ matrix obtained by removing the i th row and the j th column of the matrix A .

EXAMPLE:

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

• Let A be an $(n \times n)$ matrix whose entry $(A)_{ij} = a_{ij}$. We define the **determinant** of A as

$$\det A = |A| = \sum_{j=1}^n (-1)^{j+1} a_{1j} \det A_{1j} = \sum_{j=1}^n a_{1j} C_{1j},$$

where $C_{ij} = (-1)^{i+j} \det A_{ij}$ is referred to as the ij **cofactor** of A .

Theorem 4.15. The determinant of a square matrix A can be expressed as the cofactor expansion along any row of the matrix

$$\det A = \sum_{j=1}^n (-1)^{k+j} a_{kj} \det A_{kj} = \sum_{j=1}^n a_{kj} C_{kj} \quad \left(\begin{array}{l} \text{along the} \\ k\text{th row} \end{array} \right)$$

WARNING:

-
-

Theorem 4.16. If A is an $(n \times n)$ triangular matrix, its determinant is the product of its diagonal entries.

$$\det \begin{bmatrix} a_{11} & 0 & 0 & 0 & 0 \\ * & a_{22} & 0 & 0 & 0 \\ * & * & a_{33} & 0 & 0 \\ * & * & * & a_{44} & 0 \\ * & * & * & * & a_{55} \end{bmatrix} =$$

EXAMPLE:

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

1st row:

=

2nd row:

=

Theorem 4.17. Let A be an $(n \times n)$ matrix.

If we obtain a matrix B ,

- By adding to a row of A the multiple of another row,

$$\det B = \det A.$$

- By multiplying one row of A by λ ,

$$\det B = \lambda \det A.$$

- By interchanging two rows of A ,

$$\det B = -\det A.$$

EXAMPLE:

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

Theorem 4.18. Let A be a square matrix and U an echelon matrix obtained from A by adding multiples of rows and r row interchanges (but without multiplying any row by a scalar!).

Then,

$$\det A = \begin{cases} 0 & \text{if } A \text{ is not invertible} \\ (-1)^r \cdot (\text{product of the pivots}) & \text{if } A \text{ is invertible} \end{cases}$$

Proof:

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Note: This would add a new statement to theorem 4.11:

19. The determinant of A is nonzero.

WARNING: In general,

$$A \sim B \not\Rightarrow \det A = \det B.$$

Check theorem 4.17!

WARNING: In general,

$$\det(A + B) \neq \det A + \det B.$$

EXAMPLE: *If it was true, all determinants would be zero:*

$$\det \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \det \left(\begin{bmatrix} a & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 & b \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ c & 0 \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 0 & d \end{bmatrix} \right)$$

Theorem 4.19. If A and B are square matrices,

$$\det(AB) = \det A \det B.$$

Theorem 4.20. If A is a square matrix,

$$|A^T| = |A| \quad \text{and} \quad |A^*| = \overline{|A|}$$

Proof:

- For elementary matrices, it's easy to see that $|E| = |E^T|$.
- If we obtain an echelon form of a matrix A :

Leading to

- Now, as U is a triangular matrix, $|U^T| = |U|$ and, consequently

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