

# Lecture Notes 1: Matrix Algebra

## Part B: Determinants and Inverses

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# Outline

## Special Matrices

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# Square Matrices

A **square matrix** has an equal number of rows and columns, this number being called its **dimension**.

The (principal, or main) **diagonal** of a square matrix  $\mathbf{A} = (a_{ij})_{n \times n}$  of dimension  $n$  is the list  $(a_{ii})_{i=1}^n = (a_{11}, a_{22}, \dots, a_{nn})$  of its  $n$  **diagonal elements**.

The other elements  $a_{ij}$  with  $i \neq j$  are the **off-diagonal elements**.

A square matrix is often expressed in the form

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix}$$

with some extra dots along the diagonal.

# Symmetric Matrices

## Definition

A square matrix  $\mathbf{A}$  is **symmetric** just in case it is equal to its transpose — i.e., if  $\mathbf{A}^T = \mathbf{A}$ .

## Example

The product of two symmetric matrices need not be symmetric.

Using again our example of non-commuting  $2 \times 2$  matrices, here are two examples

where the product of two symmetric matrices is asymmetric:

$$\blacktriangleright \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix};$$

$$\blacktriangleright \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}$$

## Two Exercises with Symmetric Matrices

### Exercise

Let  $\mathbf{x}$  be a column  $n$ -vector.

1. Find the dimensions of  $\mathbf{x}^\top \mathbf{x}$  and of  $\mathbf{x}\mathbf{x}^\top$ .
2. Show that one is a non-negative number which is positive unless  $\mathbf{x} = \mathbf{0}$ , and that the other is an  $n \times n$  symmetric matrix.

### Exercise

Let  $\mathbf{A}$  be an  $m \times n$ -matrix.

1. Find the dimensions of  $\mathbf{A}^\top \mathbf{A}$  and of  $\mathbf{A}\mathbf{A}^\top$ .
2. Show that both  $\mathbf{A}^\top \mathbf{A}$  and  $\mathbf{A}\mathbf{A}^\top$  are symmetric matrices.
3. Show that  $m = n$  is a necessary condition for  $\mathbf{A}^\top \mathbf{A} = \mathbf{A}\mathbf{A}^\top$ .
4. Show that  $m = n$  with  $\mathbf{A}$  symmetric is a sufficient condition for  $\mathbf{A}^\top \mathbf{A} = \mathbf{A}\mathbf{A}^\top$ .

## Diagonal Matrices

A square matrix  $\mathbf{A} = (a_{ij})^{n \times n}$  is **diagonal** just in case all of its off diagonal elements are 0 — i.e.,  $i \neq j \implies a_{ij} = 0$ .

A diagonal matrix of dimension  $n$  can be written in the form

$$\mathbf{D} = \begin{pmatrix} d_1 & 0 & 0 & \dots & 0 \\ 0 & d_2 & 0 & \dots & 0 \\ 0 & 0 & d_3 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & d_n \end{pmatrix} = \mathbf{diag}(d_1, d_2, d_3, \dots, d_n) = \mathbf{diag} \mathbf{d}$$

where the  $n$ -vector  $\mathbf{d} = (d_1, d_2, d_3, \dots, d_n) = (d_i)_{i=1}^n$  consists of the diagonal elements of  $\mathbf{D}$ .

Note that  $\mathbf{diag} \mathbf{d} = (d_{ij})_{n \times n}$  where each  $d_{ij} = \delta_{ij} d_{ii} = \delta_{ij} d_{jj}$ .

Obviously, any diagonal matrix is symmetric.

# Multiplying by Diagonal Matrices

## Example

Let  $\mathbf{D}$  be a diagonal matrix of dimension  $n$ .

Suppose that  $\mathbf{A}$  and  $\mathbf{B}$  are  $m \times n$  and  $n \times m$  matrices, respectively.

Then  $\mathbf{E} := \mathbf{AD}$  and  $\mathbf{F} := \mathbf{DB}$  are well defined matrices of dimensions  $m \times n$  and  $n \times m$ , respectively.

By the law of matrix multiplication, their elements are

$$e_{ij} = \sum_{k=1}^n a_{ik} \delta_{kj} d_{jj} = a_{ij} d_{jj} \text{ and } f_{ij} = \sum_{k=1}^n \delta_{ik} d_{ii} b_{kj} = d_{ii} b_{ij}$$

Thus, **post**-multiplying  $\mathbf{A}$  by  $\mathbf{D}$  is the **column** operation of simultaneously multiplying every column  $\mathbf{a}_j$  of  $\mathbf{A}$  by its matching diagonal element  $d_{jj}$ .

Similarly, **pre**-multiplying  $\mathbf{B}$  by  $\mathbf{D}$  is the **row** operation of simultaneously multiplying every row  $\mathbf{b}_i^T$  of  $\mathbf{B}$  by its matching diagonal element  $d_{ii}$ .

## Two Exercises with Diagonal Matrices

### Exercise

Let  $\mathbf{D}$  be a diagonal matrix of dimension  $n$ .

Give conditions that are both necessary and sufficient for each of the following:

1.  $\mathbf{AD} = \mathbf{A}$  for every  $m \times n$  matrix  $\mathbf{A}$ ;
2.  $\mathbf{DB} = \mathbf{B}$  for every  $n \times m$  matrix  $\mathbf{B}$ .

### Exercise

Let  $\mathbf{D}$  be a diagonal matrix of dimension  $n$ , and  $\mathbf{C}$  any  $n \times n$  matrix.

An earlier example shows that one can have  $\mathbf{CD} \neq \mathbf{DC}$  even if  $n = 2$ .

1. Show that  $\mathbf{C}$  being diagonal is a sufficient condition for  $\mathbf{CD} = \mathbf{DC}$ .
2. Is this condition necessary?



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# The Identity Matrix

The **identity matrix** of dimension  $n$  is the diagonal matrix

$$\mathbf{I}_n = \mathbf{diag}(1, 1, \dots, 1)$$

whose  $n$  diagonal elements are all equal to 1.

Equivalently, it is the  $n \times n$ -matrix  $\mathbf{A} = (a_{ij})^{n \times n}$

whose elements are all given by  $a_{ij} = \delta_{ij}$

for the Kronecker delta function  $(i, j) \mapsto \delta_{ij}$

defined on  $\{1, 2, \dots, n\}^2$ .

## Exercise

Given any  $m \times n$  matrix  $\mathbf{A}$ , verify that  $\mathbf{I}_m \mathbf{A} = \mathbf{A} \mathbf{I}_n = \mathbf{A}$ .

# Uniqueness of the Identity Matrix

## Exercise

Suppose that the two  $n \times n$  matrices  $\mathbf{X}$  and  $\mathbf{Y}$  respectively satisfy:

1.  $\mathbf{AX} = \mathbf{A}$  for every  $m \times n$  matrix  $\mathbf{A}$ ;
2.  $\mathbf{YB} = \mathbf{B}$  for every  $n \times m$  matrix  $\mathbf{B}$ .

Prove that  $\mathbf{X} = \mathbf{Y} = \mathbf{I}_n$ .

(Hint: Consider each of the  $mn$  different cases where  $\mathbf{A}$  (resp.  $\mathbf{B}$ ) has exactly one non-zero element that is equal to 1.)

The results of the last two exercises together serve to prove:

## Theorem

The identity matrix  $\mathbf{I}_n$  is the unique  $n \times n$ -matrix such that:

- ▶  $\mathbf{I}_n \mathbf{B} = \mathbf{B}$  for each  $n \times m$  matrix  $\mathbf{B}$ ;
- ▶  $\mathbf{A} \mathbf{I}_n = \mathbf{A}$  for each  $m \times n$  matrix  $\mathbf{A}$ .

# How the Identity Matrix Earns its Name

## Remark

*The identity matrix  $\mathbf{I}_n$  earns its name because it represents a **multiplicative identity** on the “algebra” of all  $n \times n$  matrices.*

*That is,  $\mathbf{I}_n$  is the unique  $n \times n$ -matrix with the property that  $\mathbf{I}_n \mathbf{A} = \mathbf{A} \mathbf{I}_n = \mathbf{A}$  for every  $n \times n$ -matrix  $\mathbf{A}$ .*

Typical notation suppresses the subscript  $n$  in  $\mathbf{I}_n$  that indicates the dimension of the identity matrix.

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# Left and Right Inverse Matrices

## Definition

Let  $\mathbf{A}$  denote any  $n \times n$  matrix.

1. The  $n \times n$  matrix  $\mathbf{X}$  is a **left inverse** of  $\mathbf{A}$  just in case  $\mathbf{XA} = \mathbf{I}_n$ .
2. The  $n \times n$  matrix  $\mathbf{Y}$  is a **right inverse** of  $\mathbf{A}$  just in case  $\mathbf{AY} = \mathbf{I}_n$ .
3. The  $n \times n$  matrix  $\mathbf{Z}$  is an **inverse** of  $\mathbf{A}$  just in case it is both a left and a right inverse — i.e.,  $\mathbf{ZA} = \mathbf{AZ} = \mathbf{I}_n$ .

# The Unique Inverse Matrix

## Theorem

Suppose that the  $n \times n$  matrix  $\mathbf{A}$  has both a left and a right inverse. Then both left and right inverses are unique, and both are equal to a unique *inverse matrix* denoted by  $\mathbf{A}^{-1}$ .

## Proof.

If  $\mathbf{XA} = \mathbf{AY} = \mathbf{I}$ , then  $\mathbf{XAY} = \mathbf{XI} = \mathbf{X}$  and  $\mathbf{XAY} = \mathbf{IY} = \mathbf{Y}$ , implying that  $\mathbf{X} = \mathbf{XAY} = \mathbf{Y}$ .

Now, if  $\tilde{\mathbf{X}}$  is any alternative left inverse, then  $\tilde{\mathbf{X}}\mathbf{A} = \mathbf{I}$  and so  $\tilde{\mathbf{X}} = \tilde{\mathbf{X}}\mathbf{AY} = \mathbf{Y} = \mathbf{X}$ .

Similarly, if  $\tilde{\mathbf{Y}}$  is any alternative right inverse, then  $\mathbf{A}\tilde{\mathbf{Y}} = \mathbf{I}$  and so  $\tilde{\mathbf{Y}} = \mathbf{XA}\tilde{\mathbf{Y}} = \mathbf{X} = \mathbf{Y}$ .

It follows that  $\tilde{\mathbf{X}} = \mathbf{X} = \mathbf{Y} = \tilde{\mathbf{Y}}$ , so we can define  $\mathbf{A}^{-1}$  as the unique common value of all these four matrices. □

**Big question:** when does the inverse exist?

**Answer:** if and only if the *determinant* is non-zero.

# Rule for Inverting Products

## Theorem

Suppose that  $\mathbf{A}$  and  $\mathbf{B}$  are two invertible  $n \times n$  matrices.

Then the inverse of the matrix product  $\mathbf{AB}$  exists, and is the reverse product  $\mathbf{B}^{-1}\mathbf{A}^{-1}$  of the inverses.

## Proof.

Using the associative law for matrix multiplication repeatedly gives:

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

and

$$(\mathbf{AB})(\mathbf{B}^{-1}\mathbf{A}^{-1}) = \mathbf{A}(\mathbf{BB}^{-1})\mathbf{A}^{-1} = \mathbf{A}(\mathbf{I})\mathbf{A}^{-1} = (\mathbf{AI})\mathbf{A}^{-1} = \mathbf{AA}^{-1} = \mathbf{I}.$$

These equations confirm that  $\mathbf{X} := \mathbf{B}^{-1}\mathbf{A}^{-1}$  is the unique matrix satisfying the double equality  $(\mathbf{AB})\mathbf{X} = \mathbf{X}(\mathbf{AB}) = \mathbf{I}$ . □



## Rule for Inverting Chain Products and Transposes

### Exercise

Prove that, if  $\mathbf{A}$ ,  $\mathbf{B}$  and  $\mathbf{C}$  are three invertible  $n \times n$  matrices, then  $(\mathbf{ABC})^{-1} = \mathbf{C}^{-1}\mathbf{B}^{-1}\mathbf{A}^{-1}$ .

Then use mathematical induction to extend the rule for inverting any product  $\mathbf{BC}$  in order to find the inverse of the product  $\mathbf{A}_1\mathbf{A}_2 \cdots \mathbf{A}_k$  of any finite chain of invertible  $n \times n$  matrices.

### Theorem

Suppose that  $\mathbf{A}$  is an invertible  $n \times n$  matrix.

Then the inverse  $(\mathbf{A}^\top)^{-1}$  of its transpose is  $(\mathbf{A}^{-1})^\top$ , the transpose of its inverse.

### Proof.

By the rule for transposing products, one has

$$\mathbf{A}^\top(\mathbf{A}^{-1})^\top = (\mathbf{A}^{-1}\mathbf{A})^\top = \mathbf{I}^\top = \mathbf{I}$$



# Orthogonal and Orthonormal Sets of Vectors

## Definition

A set of  $k$  vectors  $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\} \subset \mathbb{R}^n$  is said to be:

- ▶ **pairwise orthogonal** just in case  $\mathbf{x}_i \cdot \mathbf{x}_j = 0$  whenever  $j \neq i$ ;
- ▶ **orthonormal** just in case, in addition, each  $\|\mathbf{x}_i\| = 1$   
— i.e., all  $k$  elements of the set are vectors of unit length.

The set of  $k$  vectors  $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\} \subset \mathbb{R}^n$  is orthonormal just in case  $\mathbf{x}_i \cdot \mathbf{x}_j = \delta_{ij}$  for all pairs  $i, j \in \{1, 2, \dots, k\}$ .

# Orthogonal Matrices

## Definition

Any  $n \times n$  matrix is **orthogonal** just in case its  $n$  columns form an orthonormal set.

## Theorem

*Given any  $n \times n$  matrix  $\mathbf{P}$ , the following are equivalent:*

1.  $\mathbf{P}$  is orthogonal;
2.  $\mathbf{P}\mathbf{P}^\top = \mathbf{P}^\top\mathbf{P} = \mathbf{I}$ ;
3.  $\mathbf{P}^{-1} = \mathbf{P}^\top$ ;
4.  $\mathbf{P}^\top$  is orthogonal.

The proof follows from the definitions, and is left as an exercise.

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## Partitioned Matrices: Definition

A **partitioned matrix** is a rectangular array of different matrices.

### Example

Consider the  $(m + \ell) \times (n + k)$  matrix

$$\begin{pmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{pmatrix} = \begin{pmatrix} \mathbf{A}_{m \times n} & \mathbf{B}_{m \times k} \\ \mathbf{C}_{\ell \times n} & \mathbf{D}_{\ell \times k} \end{pmatrix}$$

where, as indicated, the four submatrices **A**, **B**, **C**, **D** are of dimension  $m \times n$ ,  $m \times k$ ,  $\ell \times n$  and  $\ell \times k$  respectively.

**Note:** Here matrix **D** may not be diagonal, or even square.

For any scalar  $\alpha \in \mathbb{R}$ ,

the scalar multiple of a partitioned matrix is

$$\alpha \begin{pmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{pmatrix} = \begin{pmatrix} \alpha \mathbf{A} & \alpha \mathbf{B} \\ \alpha \mathbf{C} & \alpha \mathbf{D} \end{pmatrix}$$

## Partitioned Matrices: Addition

Suppose the two partitioned matrices

$$\begin{pmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} \mathbf{E} & \mathbf{F} \\ \mathbf{G} & \mathbf{H} \end{pmatrix}$$

have the property that the following four pairs of corresponding matrices have equal dimensions:

(i) **A** and **E**; (ii) **B** and **F**; (iii) **C** and **G**; (iv) **D** and **H**.

Then the sum of the two matrices is

$$\begin{pmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{pmatrix} + \begin{pmatrix} \mathbf{E} & \mathbf{F} \\ \mathbf{G} & \mathbf{H} \end{pmatrix} = \begin{pmatrix} \mathbf{A} + \mathbf{E} & \mathbf{B} + \mathbf{F} \\ \mathbf{C} + \mathbf{G} & \mathbf{D} + \mathbf{H} \end{pmatrix}$$

## Partitioned Matrices: Multiplication

Suppose that the two partitioned matrices

$$\begin{pmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} \mathbf{E} & \mathbf{F} \\ \mathbf{G} & \mathbf{H} \end{pmatrix}$$

along with all the relevant pairs of their sub-matrices, are **compatible for multiplication**.

Then their product is defined as

$$\begin{pmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{pmatrix} \begin{pmatrix} \mathbf{E} & \mathbf{F} \\ \mathbf{G} & \mathbf{H} \end{pmatrix} = \begin{pmatrix} \mathbf{AE} + \mathbf{BG} & \mathbf{AF} + \mathbf{BH} \\ \mathbf{CE} + \mathbf{DG} & \mathbf{CF} + \mathbf{DH} \end{pmatrix}$$

This extends the usual multiplication rule for matrices: multiply the **rows** of sub-matrices in the first partitioned matrix by the **columns** of sub-matrices in the second partitioned matrix.

## Transposes and Some Special Matrices

The rule for transposing a partitioned matrix is

$$\begin{pmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{pmatrix}^T = \begin{pmatrix} \mathbf{A}^T & \mathbf{C}^T \\ \mathbf{B}^T & \mathbf{D}^T \end{pmatrix}$$

So the original matrix is symmetric

iff  $\mathbf{A} = \mathbf{A}^T$ ,  $\mathbf{D} = \mathbf{D}^T$ , and  $\mathbf{B} = \mathbf{C}^T \iff \mathbf{C} = \mathbf{B}^T$ .

It is diagonal iff  $\mathbf{A}, \mathbf{D}$  are both diagonal,  
while also  $\mathbf{B} = \mathbf{0}$  and  $\mathbf{C} = \mathbf{0}$ .

The identity matrix is diagonal with  $\mathbf{A} = \mathbf{I}$ ,  $\mathbf{D} = \mathbf{I}$ ,  
possibly identity matrices of different dimensions.



## Partitioned Matrices: Inverses, I

For an  $(m + n) \times (m + n)$  partitioned matrix to have an inverse, the equation

$$\begin{pmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{pmatrix} \begin{pmatrix} \mathbf{E} & \mathbf{F} \\ \mathbf{G} & \mathbf{H} \end{pmatrix} = \begin{pmatrix} \mathbf{AE} + \mathbf{BG} & \mathbf{AF} + \mathbf{BH} \\ \mathbf{CE} + \mathbf{DG} & \mathbf{CF} + \mathbf{DH} \end{pmatrix} = \begin{pmatrix} \mathbf{I}_m & \mathbf{0}_{m \times n} \\ \mathbf{0}_{n \times m} & \mathbf{I}_n \end{pmatrix}$$

should have a solution for the matrices  $\mathbf{E}, \mathbf{F}, \mathbf{G}, \mathbf{H}$ , given  $\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D}$ .

Assuming that the  $m \times m$  matrix  $\mathbf{A}$  has an inverse, we can:

1. construct new first  $m$  equations  
by premultiplying the old ones by  $\mathbf{A}^{-1}$ ;
2. construct new second  $n$  equations by:
  - ▶ premultiplying the new first  $m$  equations by the  $n \times m$  matrix  $\mathbf{C}$ ;
  - ▶ then subtracting this product from the old second  $n$  equations.

The result is

$$\begin{pmatrix} \mathbf{I}_m & \mathbf{A}^{-1}\mathbf{B} \\ \mathbf{0}_{n \times m} & \mathbf{D} - \mathbf{CA}^{-1}\mathbf{B} \end{pmatrix} \begin{pmatrix} \mathbf{E} & \mathbf{F} \\ \mathbf{G} & \mathbf{H} \end{pmatrix} = \begin{pmatrix} \mathbf{A}^{-1} & \mathbf{0}_{m \times n} \\ -\mathbf{CA}^{-1} & \mathbf{I}_n \end{pmatrix}$$

## Partitioned Matrices: Inverses, II

For the next step,

assume the  $n \times n$  matrix  $\mathbf{X} := \mathbf{D} - \mathbf{CA}^{-1}\mathbf{B}$

also has an inverse  $\mathbf{X}^{-1} = (\mathbf{D} - \mathbf{CA}^{-1}\mathbf{B})^{-1}$ .

$$\text{Given } \begin{pmatrix} \mathbf{I}_m & \mathbf{A}^{-1}\mathbf{B} \\ \mathbf{0}_{n \times m} & \mathbf{D} - \mathbf{CA}^{-1}\mathbf{B} \end{pmatrix} \begin{pmatrix} \mathbf{E} & \mathbf{F} \\ \mathbf{G} & \mathbf{H} \end{pmatrix} = \begin{pmatrix} \mathbf{A}^{-1} & \mathbf{0}_{m \times n} \\ -\mathbf{CA}^{-1} & \mathbf{I}_n \end{pmatrix},$$

we first premultiply the last  $n$  equations by  $\mathbf{X}^{-1}$  to get

$$\begin{pmatrix} \mathbf{I}_m & \mathbf{A}^{-1}\mathbf{B} \\ \mathbf{0}_{n \times m} & \mathbf{I}_n \end{pmatrix} \begin{pmatrix} \mathbf{E} & \mathbf{F} \\ \mathbf{G} & \mathbf{H} \end{pmatrix} = \begin{pmatrix} \mathbf{A}^{-1} & \mathbf{0}_{m \times n} \\ -\mathbf{X}^{-1}\mathbf{CA}^{-1} & \mathbf{X}^{-1} \end{pmatrix}$$

Next, we subtract  $\mathbf{A}^{-1}\mathbf{B}$  times the last  $n$  equations from the first  $m$  equations to obtain

$$\begin{aligned} \begin{pmatrix} \mathbf{E} & \mathbf{F} \\ \mathbf{G} & \mathbf{H} \end{pmatrix} &= \begin{pmatrix} \mathbf{I}_m & \mathbf{0}_{m \times n} \\ \mathbf{0}_{n \times m} & \mathbf{I}_n \end{pmatrix} \begin{pmatrix} \mathbf{E} & \mathbf{F} \\ \mathbf{G} & \mathbf{H} \end{pmatrix} \\ &= \begin{pmatrix} \mathbf{A}^{-1} + \mathbf{A}^{-1}\mathbf{B}\mathbf{X}^{-1}\mathbf{CA}^{-1} & -\mathbf{A}^{-1}\mathbf{B}\mathbf{X}^{-1} \\ -\mathbf{X}^{-1}\mathbf{CA}^{-1} & \mathbf{X}^{-1} \end{pmatrix} \end{aligned}$$

# Final Exercises

## Exercise

1. Assume that  $\mathbf{A}^{-1}$  and  $\mathbf{X}^{-1} = (\mathbf{D} - \mathbf{CA}^{-1}\mathbf{B})^{-1}$  exist.

$$\text{Given } \mathbf{Z} := \begin{pmatrix} \mathbf{A}^{-1} + \mathbf{A}^{-1}\mathbf{B}\mathbf{X}^{-1}\mathbf{C}\mathbf{A}^{-1} & -\mathbf{A}^{-1}\mathbf{B}\mathbf{X}^{-1} \\ -\mathbf{X}^{-1}\mathbf{C}\mathbf{A}^{-1} & \mathbf{X}^{-1} \end{pmatrix},$$

use direct multiplication twice in order to verify that

$$\begin{pmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{pmatrix} \mathbf{Z} = \mathbf{Z} \begin{pmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{pmatrix} = \begin{pmatrix} \mathbf{I}_m & \mathbf{0}_{m \times n} \\ \mathbf{0}_{n \times m} & \mathbf{I}_n \end{pmatrix}$$

2. Let  $\mathbf{A}$  be any invertible  $m \times m$  matrix.

Show that the bordered  $(m+1) \times (m+1)$  matrix  $\begin{pmatrix} \mathbf{A} & \mathbf{b} \\ \mathbf{c}^\top & d \end{pmatrix}$

is invertible provided that  $d \neq \mathbf{c}^\top \mathbf{A}^{-1} \mathbf{b}$ ,  
and find its inverse in this case.

# Partitioned Matrices: Extension

## Exercise

Suppose that the two partitioned matrices

$$\mathbf{A} = (\mathbf{A}_{ij})^{k \times \ell} \quad \text{and} \quad \mathbf{B} = (\mathbf{B}_{ij})^{k \times \ell}$$

are both  $k \times \ell$  arrays of respective  $m_i \times n_j$  matrices  $\mathbf{A}_{ij}, \mathbf{B}_{ij}$ , for  $i = 1, 2, \dots, k$  and  $j = 1, 2, \dots, \ell$ .

1. Under what conditions can the product  $\mathbf{AB}$  be defined as a  $k \times \ell$  array of matrices?
2. Under what conditions can the product  $\mathbf{BA}$  be defined as a  $k \times \ell$  array of matrices?
3. When either  $\mathbf{AB}$  or  $\mathbf{BA}$  can be so defined, give a formula for its product, using summation notation.
4. Express  $\mathbf{A}^\top$  as a partitioned matrix.
5. Under what conditions is the matrix  $\mathbf{A}$  symmetric?

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# Permutations

## Definition

Given  $\mathbb{N}_n = \{1, \dots, n\}$  for any  $n \in \mathbb{N}$  with  $n \geq 2$ , a **permutation** of  $\mathbb{N}_n$  is a **bijective** mapping  $\mathbb{N}_n \ni i \mapsto \pi(i) \in \mathbb{N}_n$ .

That is, the mapping  $\mathbb{N}_n \ni i \mapsto \pi(i) \in \mathbb{N}_n$  is both:

1. a **surjection**, or mapping of  $\mathbb{N}_n$  **onto**  $\mathbb{N}_n$ , in the sense that the range set satisfies  $\pi(\mathbb{N}_n) := \{j \in \mathbb{N}_n \mid \exists i \in \mathbb{N}_n : j = \pi(i)\} = \mathbb{N}_n$ ;
2. an **injection**, or a **one to one** mapping, in the sense that  $\pi(i) = \pi(j) \implies i = j$  or, equivalently,  $i \neq j \implies \pi(i) \neq \pi(j)$ .

## Exercise

*Prove that the mapping  $\mathbb{N}_n \ni i \mapsto f(i) \in \mathbb{N}_n$  is a bijection, and so a permutation, if and only if its range set  $f(\mathbb{N}_n) := \{j \in \mathbb{N}_n \mid \exists i \in \mathbb{N}_n : j = f(i)\}$  has cardinality  $\#f(\mathbb{N}_n) = \#\mathbb{N}_n = n$ .*

# Products of Permutations

## Definition

The **product**  $\pi \circ \rho$  of two permutations  $\pi, \rho \in \Pi_n$  is the composition mapping  $\mathbb{N}_n \ni i \mapsto (\pi \circ \rho)(i) := \pi[\rho(i)] \in \mathbb{N}_n$ .

## Exercise

*Prove that the product  $\pi \circ \rho$  of any two permutations  $\pi, \rho \in \Pi_n$  is a permutation.*

*Hint: Show that  $\#(\pi \circ \rho)(\mathbb{N}_n) = \#\rho(\mathbb{N}_n) = \#\mathbb{N}_n = n$ .*

## Example

1. If you shuffle a pack of 52 playing cards once, without dropping any on the floor, the result will be a permutation  $\pi$  of the cards.
2. If you shuffle the same pack a second time, the result will be a new permutation  $\rho$  of the shuffled cards.
3. Overall, the result of shuffling the cards twice will be the single permutation  $\rho \circ \pi$ .

# Finite Permutation Groups

## Definition

Given any  $n \in \mathbb{N}$ , the family  $\Pi_n$  of all permutations of  $\mathbb{N}_n$  includes:

- ▶ the **identity** permutation  $\iota$  defined by  $\iota(h) = h$  for all  $h \in \mathbb{N}_n$ ;
- ▶ because the mapping  $\mathbb{N}_n \ni i \mapsto f(i) \in \mathbb{N}_n$  is bijective, for each  $\pi \in \Pi_n$ , a unique **inverse** permutation  $\pi^{-1} \in \Pi_n$  satisfying  $\pi^{-1} \circ \pi = \pi \circ \pi^{-1} = \iota$ .

## Definition

The **associative law for functions** says that, given any three functions  $h : X \rightarrow Y$ ,  $g : Y \rightarrow Z$  and  $f : Z \rightarrow W$ , the **composite** function  $f \circ g \circ h : X \rightarrow W$  satisfies

$$(f \circ g \circ h)(x) \equiv f(g(h(x))) \equiv [(f \circ g) \circ h](x) \equiv [f \circ (g \circ h)](x)$$

## Exercise

Given any  $n \in \mathbb{N}$ , show that  $(\Pi_n, \pi, \iota)$  is an algebraic **group** — i.e., the group operation  $(\pi, \rho) \mapsto \pi \circ \rho$  is well-defined, associative, with  $\iota$  as the unit, and an inverse  $\pi^{-1}$  for every  $\pi \in \Pi_n$ .



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# Transpositions

## Definition

For each disjoint pair  $k, \ell \in \{1, 2, \dots, n\}$ , the **transposition mapping**  $i \mapsto \tau_{k\ell}(i)$  on  $\{1, 2, \dots, n\}$  is the permutation defined by

$$\tau_{k\ell}(i) := \begin{cases} \ell & \text{if } i = k; \\ k & \text{if } i = \ell; \\ i & \text{otherwise;} \end{cases}$$

That is,  $\tau_{k\ell}$  transposes the order of  $k$  and  $\ell$ , leaving all  $i \notin \{k, \ell\}$  unchanged. □

Evidently  $\tau_{k\ell} = \tau_{\ell k}$  and  $\tau_{k\ell} \circ \tau_{\ell k} = \iota$ , the identity permutation, and so  $\tau \circ \tau = \iota$  for every transposition  $\tau$ .

# Transposition is Not Commutative

Any  $(j_1, j_2, \dots, j_n) \in \mathbb{N}_n^n$  whose components are all different corresponds to a unique permutation, denoted by  $\pi^{j_1 j_2 \dots j_n} \in \Pi_n$ , that satisfies  $\pi(i) = j_i$  for all  $i \in \mathbb{N}_n^n$ .

## Example

Two transpositions defined on a set containing more than two elements **may not commute** because, for example,

$$\tau_{12} \circ \tau_{23} = \pi^{231} \neq \tau_{23} \circ \tau_{12} = \pi^{312}$$

# Permutations are Products of Transpositions

## Theorem

*Any permutation  $\pi \in \Pi_n$  on  $\mathbb{N}_n := \{1, 2, \dots, n\}$  is the product of at most  $n - 1$  transpositions.*

We will prove the result by induction on  $n$ .

As the induction hypothesis,

suppose the result holds for permutations on  $\mathbb{N}_{n-1}$ .

Any permutation  $\pi$  on  $\mathbb{N}_2 := \{1, 2\}$  is either the identity, or the transposition  $\tau_{12}$ , so the result holds for  $n = 2$ .

## Proof of Induction Step

For general  $n$ , let  $j := \pi^{-1}(n)$  denote the element that  $\pi$  moves to the end.

By construction, the permutation  $\pi \circ \tau_{jn}$  must satisfy  $\pi \circ \tau_{jn}(n) = \pi(\tau_{jn}(n)) = \pi(j) = n$ .

So the restriction  $\tilde{\pi}$  of  $\pi \circ \tau_{jn}$  to  $\mathbb{N}_{n-1}$  is a permutation on  $\mathbb{N}_{n-1}$ .

By the induction hypothesis, for all  $k \in \mathbb{N}_{n-1}$ ,

there exist transpositions  $\tau^1, \tau^2, \dots, \tau^q$

such that  $\tilde{\pi}(k) = (\pi \circ \tau_{jn})(k) = (\tau^1 \circ \tau^2 \circ \dots \circ \tau^q)(k)$

where  $q \leq n - 2$  is the number of transpositions in the product.

For  $p = 1, \dots, q$ , because  $\tau^p$  interchanges only elements of  $\mathbb{N}_{n-1}$ , one can extend its domain to include  $n$  by letting  $\tau^p(n) = n$ .

Then  $(\pi \circ \tau_{jn})(k) = (\tau^1 \circ \tau^2 \circ \dots \circ \tau^q)(k)$  for  $k = n$  as well,

so  $\pi = (\pi \circ \tau_{jn}) \circ \tau_{jn}^{-1} = \tau^1 \circ \tau^2 \circ \dots \circ \tau^q \circ \tau_{jn}^{-1}$ .

Hence  $\pi$  is the product of at most  $q + 1 \leq n - 1$  transpositions.

This completes the proof by induction on  $n$ . □

# Adjacency Transpositions and Their Products, I

## Definition

For each  $k \in \{1, 2, \dots, n-1\}$ , the transposition  $\tau_{k,k+1}$  of element  $k$  with its successor is an **adjacency transposition**. □

## Definition

For each pair  $k, \ell \in \mathbb{N}_n$  with  $k < \ell$ , define:

1.  $\pi^{k \nearrow \ell} := \tau_{\ell-1,\ell} \circ \tau_{\ell-2,\ell-1} \circ \dots \circ \tau_{k,k+1} \in \Pi_n$   
as the composition of  $\ell - k$   
successive adjacency transpositions in order,  
starting with  $\tau_{k,k+1}$  and ending with  $\tau_{\ell-1,\ell}$ ;
2.  $\pi^{\ell \searrow k} := \tau_{k,k+1} \circ \tau_{k+1,k+2} \circ \dots \circ \tau_{\ell-1,\ell} \in \Pi_n$   
as the composition of the same  $\ell - k$   
successive adjacency transpositions in reverse order.

# Adjacency Transpositions and Their Products, II

## Exercise

For each pair  $k, \ell \in \mathbb{N}_n$  with  $k < \ell$ , prove that:

$$\blacktriangleright \pi^{k \nearrow \ell}(i) := \begin{cases} i & \text{if } i < k \text{ or } i > \ell; \\ i - 1 & \text{if } k < i \leq \ell; \\ \ell & \text{if } i = k. \end{cases}$$

$$\blacktriangleright \pi^{k \nearrow k} = \pi^{k \searrow k} = \iota$$

$\blacktriangleright \pi^{k \nearrow \ell}$  and  $\pi^{\ell \searrow k}$  are inverses

$$\blacktriangleright \pi^{k \nearrow \ell} = \pi^{1, 2, \dots, k-1, k+1, \dots, \ell-1, \ell, k, \ell+1, \dots, n}$$

$$\blacktriangleright \pi^{\ell \searrow k} = \pi^{1, 2, \dots, k-1, \ell, k, k+1, \dots, \ell-2, \ell-1, \ell+1, \dots, n} \quad \square$$

1. Note that  $\pi^{k \nearrow \ell}$  moves  $k$  up to the  $\ell$ th position, while moving each element between  $k + 1$  and  $\ell$  down by one.
2. By contrast,  $\pi^{\ell \searrow k}$  moves  $\ell$  down to the  $k$ th position, while moving each element between  $k$  and  $\ell - 1$  up by one.

# Reduction to the Product of Adjacency Transpositions

## Lemma

For each pair  $k, \ell \in \mathbb{N}_n$  with  $k < \ell$ , the transposition  $\tau_{k\ell}$  equals both  $\pi^{\ell-1} \searrow k \circ \pi^k \nearrow \ell$  and  $\pi^{k+1} \nearrow \ell \circ \pi^\ell \searrow k$ , the compositions of  $2(\ell - k) - 1$  adjacency transpositions.

## Proof.

1. As noted,  $\pi^k \nearrow \ell$  moves  $k$  up to the  $\ell$ th position, while moving each element between  $k + 1$  and  $\ell$  down by one. Then  $\pi^{\ell-1} \searrow k$  moves  $\ell$ , which  $\pi^k \nearrow \ell$  left in position  $\ell - 1$ , down to the  $k$  position, and moves  $k + 1, k + 2, \dots, \ell - 1$  up by one, back to their original positions.

This proves that  $\pi^{\ell-1} \searrow k \circ \pi^k \nearrow \ell = \tau_{k\ell}$ .

It also expresses  $\tau_{k\ell}$  as the composition of  $(\ell - k) + (\ell - 1 - k) = 2(\ell - k) - 1$  adjacency transpositions.

2. The proof that  $\pi^{k+1} \nearrow \ell \circ \pi^\ell \searrow k = \tau_{k\ell}$  is similar; details are left as an exercise. □



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# The Inversions of a Permutation

## Definition

1. Let  $\mathbb{N}_{n,2} = \{S \subseteq \mathbb{N}_n \mid \#S = 2\}$  denote the set of all (unordered) **pair subsets** of  $\mathbb{N}_n$ .
2. Obviously, if  $\{i, j\} \in \mathbb{N}_{n,2}$ , then  $i \neq j$ .
3. Given any pair  $\{i, j\} \in \mathbb{N}_{n,2}$ , define

$$i \vee j := \max\{i, j\} \quad \text{and} \quad i \wedge j := \min\{i, j\}$$

For all  $\{i, j\} \in \mathbb{N}_{n,2}$ , because  $i \neq j$ , one has  $i \vee j > i \wedge j$ .

4. Given any permutation  $\pi \in \Pi_n$ , the pair  $\{i, j\} \in \mathbb{N}_{n,2}$  is an **inversion** of  $\pi$  just in case  $\pi$  “reorders”  $\{i, j\}$  in the sense that  $\pi(i \vee j) < \pi(i \wedge j)$ .
5. Denote the set of inversions of  $\pi$  by

$$\mathfrak{N}(\pi) := \{\{i, j\} \in \mathbb{N}_{n,2} \mid \pi(i \vee j) < \pi(i \wedge j)\}$$

# The Sign of a Permutation

## Definition

1. Given any permutation  $\pi : \mathbb{N}_n \rightarrow \mathbb{N}_n$ , let  $n(\pi) := \#\mathfrak{I}(\pi) \in \mathbb{N} \cup \{0\}$  denote the number of its inversions.
2. A permutation  $\pi : \mathbb{N}_n \rightarrow \mathbb{N}_n$  is either **even** or **odd** according as  $n(\pi)$  is an even or odd number.
3. The **sign** or **signature** of a permutation  $\pi$ , is defined as  $\text{sgn}(\pi) := (-1)^{n(\pi)}$ , which is:  
(i)  $+1$  if  $\pi$  is even; (ii)  $-1$  if  $\pi$  is odd.

# The Sign of an Adjacency Transposition

## Theorem

For each  $k \in \mathbb{N}_{n-1}$ , if  $\pi$  is the adjacency transposition  $\tau_{k,k+1}$ , then  $\mathfrak{N}(\pi) = \{\{k, k+1\}\}$ , so  $n(\pi) = 1$  and  $\text{sgn}(\pi) = -1$ .

## Proof.

If  $\pi$  is the adjacency transposition  $\tau_{k,k+1}$ , then

$$\pi(i) = \begin{cases} i & \text{if } i \notin \{k, k+1\} \\ k+1 & \text{if } i = k \\ k & \text{if } i = k+1 \end{cases}$$

It is evident that  $\{k, k+1\}$  is an inversion.

Also  $\pi(i) \leq i$  for all  $i \neq k$ , and  $\pi(j) \geq j$  for all  $j \neq k+1$ .

So if  $i < j$ , then  $\pi(i) \leq i < j \leq \pi(j)$  unless  $i = k$  and  $j = k+1$ , and so  $\pi(i) > \pi(j)$  only if  $(i, j) = (k, k+1)$ .

Hence  $\mathfrak{N}(\pi) = \{\{k, k+1\}\}$ , implying that  $n(\pi) = 1$ . □

# A Multi-Part Exercise

## Exercise

Show that:

1. For each permutation  $\pi \in \Pi_n$ , one has

$$\begin{aligned}\mathfrak{N}(\pi) &:= \{ \{i, j\} \in \mathbb{N}_{n,2} \mid (i - j)[\pi(i) - \pi(j)] < 0 \} \\ &= \left\{ \{i, j\} \in \mathbb{N}_{n,2} \mid \frac{\pi(i) - \pi(j)}{i - j} < 0 \right\}\end{aligned}$$

2.  $\mathfrak{n}(\pi) = 0 \iff \pi = \iota$ , the identity permutation;
3.  $\mathfrak{n}(\pi) \leq \frac{1}{2}n(n - 1)$ , with equality if and only if  $\pi$  is the **reversal permutation** defined by  $\pi(i) = n - i + 1$  for all  $i \in \mathbb{N}_n$  — i.e.,

$$(\pi(1), \pi(2), \dots, \pi(n - 1), \pi(n)) = (n, n - 1, \dots, 2, 1)$$

**Hint:** Consider the number of ordered pairs  $(i, j) \in \mathbb{N}_n \times \mathbb{N}_n$  that satisfy  $i < j$ .

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## Double Products

Let  $\mathbf{X} = \langle x_{ij} \rangle_{(i,j) \in \mathbb{N}_n \times \mathbb{N}_n}$  denote an  $n \times n$  matrix.

We introduce the notation

$$\prod_{i>j}^n x_{ij} := \prod_{i=1}^n \prod_{j=1}^{i-1} x_{ij} := \prod_{j=1}^n \prod_{i=j+1}^n x_{ij}$$

for the product of all the elements in the lower triangular matrix  $\mathbf{L}$

with elements  $\ell_{ij} := \begin{cases} x_{ij} & \text{if } i > j \\ 0 & \text{if } i \leq j \end{cases}$

In case the matrix  $\mathbf{X}$  is symmetric, one has

$$\prod_{i>j}^n x_{ij} = \prod_{i>j}^n x_{ji} = \prod_{i<j}^n x_{ij}$$

This can be rewritten as  $\prod_{i>j}^n x_{ij} = \prod_{\{i,j\} \in \mathbb{N}_{n,2}} x_{ij}$ ,  
which is the product over all unordered pairs of elements in  $\mathbb{N}_n$ .

# Preliminary Example and Definition

## Example

For every  $n \in \mathbb{N}$ , define the double product

$$\mathbb{P}_{n,2} := \prod_{\{i,j\} \in \mathbb{N}_{n,2}} |i-j| = \prod_{i>j}^n |i-j| = \prod_{i<j}^n |i-j|$$

Then one has

$$\begin{aligned}\mathbb{P}_{n,2} &= (n-1)(n-2)^2(n-3)^3 \dots 3^{n-3} 2^{n-2} 1^{n-1} \\ &= \prod_{k=1}^{n-1} k^{n-k} \\ &= (n-1)!(n-2)!(n-3)! \dots 3!2! = \prod_{k=1}^{n-1} k!\end{aligned}$$

## Definition

For every permutation  $\pi \in \Pi_n$ , define the symmetric matrix  $\mathbf{X}^\pi$

$$\text{so that } x_{ij}^\pi := \begin{cases} \frac{\pi(i) - \pi(j)}{i - j} & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$$



# Basic Lemma

## Lemma

For every permutation  $\pi \in \Pi_n$ , one has  $\text{sgn}(\pi) = \prod_{\{i,j\} \in \mathbb{N}_{n,2}} x_{ij}^\pi$ .

## Proof.

- ▶ Because  $\pi$  is a permutation, the mapping  $\mathbb{N}_{n,2} \ni \{i,j\} \mapsto \{\pi(i), \pi(j)\} \in \mathbb{N}_{n,2}$  has inverse  $\mathbb{N}_{n,2} \ni \{i,j\} \mapsto \{\pi^{-1}(i), \pi^{-1}(j)\} \in \mathbb{N}_{n,2}$ . In fact it is a bijection between  $\mathbb{N}_{n,2}$  and itself.
- ▶ Hence  $\mathbb{P}_{n,2} := \prod_{\{i,j\} \in \mathbb{N}_{n,2}} |i - j| = \prod_{\{i,j\} \in \mathbb{N}_{n,2}} |\pi(i) - \pi(j)|$ .
- ▶ So  $\prod_{\{i,j\} \in \mathbb{N}_{n,2}} \frac{|\pi(i) - \pi(j)|}{|i - j|} = \prod_{\{i,j\} \in \mathbb{N}_{n,2}} |x_{ij}^\pi| = 1$ .
- ▶ Also  $x_{ij}^\pi = \mp 1$  according as  $\{i,j\}$  is or is not a reversal of  $\pi$ .
- ▶ It follows that  $\prod_{\{i,j\} \in \mathbb{N}_{n,2}} x_{ij}^\pi = (-1)^{n(\pi)} \prod_{\{i,j\} \in \mathbb{N}_{n,2}} |x_{ij}^\pi| = (-1)^{n(\pi)} = \text{sgn}(\pi)$

□

# The Product Rule for Signs of Permutations

## Theorem

For all permutations  $\rho, \pi \in \Pi_n$  one has  $\text{sgn}(\rho \circ \pi) = \text{sgn}(\rho) \text{sgn}(\pi)$ .

## Proof.

The basic lemma implies that

$$\begin{aligned} \frac{\text{sgn}(\rho \circ \pi)}{\text{sgn}(\pi)} &= \prod_{\{i,j\} \in \mathbb{N}_{n,2}} \frac{\rho(\pi(i)) - \rho(\pi(j))}{i - j} \prod_{\{k,\ell\} \in \mathbb{N}_{n,2}} \frac{k - \ell}{\pi(k) - \pi(\ell)} \\ &= \prod_{\{i,j\} \in \mathbb{N}_{n,2}} \frac{\rho(\pi(i)) - \rho(\pi(j))}{i - j} \prod_{\{i,j\} \in \mathbb{N}_{n,2}} \frac{i - j}{\pi(i) - \pi(j)} \end{aligned}$$

After cancelling the product  $\prod_{\{i,j\} \in \mathbb{N}_{n,2}} (i - j)$  and then replacing  $\pi(i)$  by  $k$  and  $\pi(j)$  by  $\ell$ , because  $\pi$  and  $\rho$  are permutations, one obtains

$$\frac{\text{sgn}(\rho \circ \pi)}{\text{sgn}(\pi)} = \prod_{\{k,\ell\} \in \mathbb{N}_{n,2}} \frac{\rho(k) - \rho(\ell)}{k - \ell} = \text{sgn}(\rho) \quad \square$$

# The Sign of the Inverse Permutations

## Corollary

Given any permutation  $\pi \in \Pi_n$ , one has  $\text{sgn}(\pi^{-1}) = \text{sgn}(\pi)$ .

## Proof.

Because the identity permutation satisfies  $\iota = \pi \circ \pi^{-1}$ , the product rule implies that

$$1 = \text{sgn}(\iota) = \text{sgn}(\pi \circ \pi^{-1}) = \text{sgn}(\pi) \text{sgn}(\pi^{-1})$$

Because  $\text{sgn}(\pi), \text{sgn}(\pi^{-1}) \in \{-1, 1\}$ , they must both have the same sign, and the result follows. □

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## Determinants of Order 2: Definition

Consider again the pair of linear equations

$$\begin{aligned}a_{11}x_1 + a_{12}x_2 &= b_1 \\ a_{21}x_1 + a_{22}x_2 &= b_2\end{aligned}$$

with its associated coefficient matrix

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$$

Let us define the number  $D := a_{11}a_{22} - a_{21}a_{12}$ .

We saw earlier that, provided that  $D \neq 0$ , the two simultaneous equations have a unique solution given by

$$x_1 = \frac{1}{D}(b_1a_{22} - b_2a_{12}), \quad x_2 = \frac{1}{D}(b_2a_{11} - b_1a_{21})$$

The number  $D$  is called the **determinant** of the matrix  $\mathbf{A}$ .

It is denoted by either  $\det(\mathbf{A})$ , or more concisely, by  $|\mathbf{A}|$ .

## Determinants of Order 2: Simple Rule

Thus, for any  $2 \times 2$  matrix  $\mathbf{A}$ , its determinant  $D$  is

$$|\mathbf{A}| = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{21}a_{12}$$

For this special case of **order 2** determinants, a simple rule is:

1. multiply the diagonal elements together;
2. multiply the off-diagonal elements together;
3. subtract the product of the off-diagonal elements from the product of the diagonal elements.

### Exercise

*Show that the determinant satisfies*

$$|\mathbf{A}| = a_{11}a_{22} \begin{vmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix} + a_{21}a_{12} \begin{vmatrix} 0 & 1 \\ 1 & 0 \end{vmatrix}$$

# Transposing the Rows or Columns

## Example

Consider the two  $2 \times 2$  matrices  $\mathbf{A} = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ ,  $\mathbf{T} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$ .

Note that  $\mathbf{T}$  is orthogonal.

Also, one has  $\mathbf{AT} = \begin{pmatrix} b & a \\ d & c \end{pmatrix}$  and  $\mathbf{TA} = \begin{pmatrix} c & d \\ a & b \end{pmatrix}$ .

Here  $\mathbf{T}$  is a **transposition matrix** which interchanges:  
(i) the **columns** of  $\mathbf{A}$  in  $\mathbf{AT}$ ; (ii) the **rows** of  $\mathbf{A}$  in  $\mathbf{TA}$ .

Evidently  $|\mathbf{T}| = -1$  and  $|\mathbf{TA}| = |\mathbf{AT}| = (bc - ad) = -|\mathbf{A}|$ .

So interchanging the two rows or columns of  $\mathbf{A}$  changes the sign of  $|\mathbf{A}|$ .

# Sign Adjusted Transpositions

## Example

Next, consider the following three  $2 \times 2$  matrices:

$$\mathbf{A} = \begin{pmatrix} a & b \\ c & d \end{pmatrix}, \quad \mathbf{T} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}, \quad \hat{\mathbf{T}} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$$

Note that, like  $\mathbf{T}$ , the matrix  $\hat{\mathbf{T}}$  is orthogonal.

Here one has  $\mathbf{A}\hat{\mathbf{T}} = \begin{pmatrix} b & -a \\ d & -c \end{pmatrix}$  and  $\hat{\mathbf{T}}\mathbf{A} = \begin{pmatrix} -c & -d \\ a & b \end{pmatrix}$ .

Evidently  $|\hat{\mathbf{T}}| = 1$  and  $|\hat{\mathbf{T}}\mathbf{A}| = |\mathbf{A}\hat{\mathbf{T}}| = (ad - bc) = |\mathbf{A}|$ .

The same is true of its transpose (and inverse)  $\hat{\mathbf{T}}^T = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$ .

This key property makes both  $\hat{\mathbf{T}}$  and  $\hat{\mathbf{T}}^T$  **sign adjusted** versions of the transposition matrix  $\mathbf{T}$ .



## Cramer's Rule in the $2 \times 2$ Case

Using determinant notation, the solution to the equations

$$\begin{aligned}a_{11}x_1 + a_{12}x_2 &= b_1 \\ a_{21}x_1 + a_{22}x_2 &= b_2\end{aligned}$$

can be written in the alternative form

$$x_1 = \frac{1}{D} \begin{vmatrix} b_1 & a_{12} \\ b_2 & a_{22} \end{vmatrix}, \quad x_2 = \frac{1}{D} \begin{vmatrix} a_{11} & b_1 \\ a_{21} & b_2 \end{vmatrix}$$

This accords with **Cramer's rule**,

which says that the solution to  $\mathbf{Ax} = \mathbf{b}$  is the vector  $\mathbf{x} = (x_i)_{i=1}^n$  each of whose components  $x_i$  is the fraction with:

1. denominator equal to the determinant  $D$  of the coefficient matrix  $\mathbf{A}$  (**provided**, of course, that  $D \neq 0$ );
2. numerator equal to the determinant of the matrix  $[\mathbf{A}_{-i}/\mathbf{b}]$  formed from  $\mathbf{A}$  by excluding its  $i$ th column, then replacing it with the  $\mathbf{b}$  vector of right-hand side elements, while keeping all the columns in their original order.

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## Determinants of Order 3: Definition

Determinants of order 3 can be calculated from those of order 2 according to the formula

$$\begin{aligned} |\mathbf{A}| &= a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix} \\ &= \sum_{j=1}^3 (-1)^{1+j} a_{1j} |\mathbf{C}_{1j}| \end{aligned}$$

where, for  $j = 1, 2, 3$ , the  $2 \times 2$  matrix  $\mathbf{C}_{1j}$  is the  $(1, j)$ -**cofactor** obtained by removing both row 1 and column  $j$  from the matrix  $\mathbf{A}$ .

The result is the following sum

$$\begin{aligned} |\mathbf{A}| &= a_{11}a_{22}a_{33} - a_{11}a_{23}a_{32} + a_{12}a_{23}a_{31} \\ &\quad - a_{12}a_{21}a_{33} + a_{13}a_{21}a_{32} - a_{13}a_{22}a_{31} \end{aligned}$$

of  $3! = 6$  terms, each the product of 3 elements chosen so that each row and each column is represented just once.

## Determinants of Order 3: Cofactor Expansion

The determinant expansion

$$\begin{aligned} |\mathbf{A}| = & a_{11}a_{22}a_{33} - a_{11}a_{23}a_{32} + a_{12}a_{23}a_{31} \\ & - a_{12}a_{21}a_{33} + a_{13}a_{21}a_{32} - a_{13}a_{22}a_{31} \end{aligned}$$

is very symmetric, suggesting (correctly)  
that the cofactor expansion **along the first row** ( $a_{11}, a_{12}, a_{13}$ )

$$|\mathbf{A}| = \sum_{j=1}^3 (-1)^{1+j} a_{1j} |\mathbf{C}_{1j}|$$

gives the same answer as the other cofactor expansions

$$|\mathbf{A}| = \sum_{j=1}^3 (-1)^{r+j} a_{rj} |\mathbf{C}_{rj}| = \sum_{i=1}^3 (-1)^{i+s} a_{is} |\mathbf{C}_{is}|$$

along, respectively:

- ▶ **the  $r$ th row** ( $a_{r1}, a_{r2}, a_{r3}$ )
- ▶ **the  $s$ th column** ( $a_{1s}, a_{2s}, a_{3s}$ )

## Determinants of Order 3: Alternative Expressions

One way of condensing the notation

$$|\mathbf{A}| = a_{11}a_{22}a_{33} - a_{11}a_{23}a_{32} + a_{12}a_{23}a_{31} \\ - a_{12}a_{21}a_{33} + a_{13}a_{21}a_{32} - a_{13}a_{22}a_{31}$$

is to reduce it to  $|\mathbf{A}| = \sum_{\pi \in \Pi_3} \text{sgn}(\pi) \prod_{i=1}^3 a_{i\pi(i)}$   
for the sign function  $\Pi_3 \ni \pi \mapsto \text{sgn}(\pi) \in \{-1, +1\}$ .

The six values of  $\text{sgn}(\pi)$  can be read off as

$$\begin{aligned} \text{sgn}(\pi^{123}) &= +1; & \text{sgn}(\pi^{132}) &= -1; & \text{sgn}(\pi^{231}) &= +1; \\ \text{sgn}(\pi^{213}) &= -1; & \text{sgn}(\pi^{312}) &= +1; & \text{sgn}(\pi^{321}) &= -1. \end{aligned}$$

### Exercise

Verify these values for each of the six  $\pi \in \Pi_3$  by:

1. calculating the number of inversions directly;
2. expressing each  $\pi$  as the product of transpositions, and then counting these.

## Sarrus's Rule: Diagram

An alternative way to evaluate determinants **only** of order 3 is to add two new columns that repeat the first and second columns:

$$\begin{array}{ccccc} a_{11} & a_{12} & a_{13} & a_{11} & a_{12} \\ a_{21} & a_{22} & a_{23} & a_{21} & a_{22} \\ a_{31} & a_{32} & a_{33} & a_{31} & a_{32} \end{array}$$

Then add lines/arrows going up to the right or down to the right, as shown below

$$\begin{array}{cccccc} a_{11} & & a_{12} & & a_{13} & & a_{11} & & a_{12} \\ & \searrow & & \nearrow & & \nearrow & & \nearrow & \\ a_{21} & & a_{22} & & a_{23} & & a_{21} & & a_{22} \\ & \nearrow & & \searrow & & \searrow & & \searrow & \\ a_{31} & & a_{32} & & a_{33} & & a_{31} & & a_{32} \end{array}$$

Note that some pairs of arrows in the middle cross each other.

## Sarrus's Rule Defined

Now:

1. multiply along the three lines falling to the right, then sum these three products, to obtain

$$a_{11} a_{22} a_{33} + a_{12} a_{23} a_{31} + a_{13} a_{21} a_{32}$$

2. multiply along the three lines rising to the right, then sum these three products, giving the sum a minus sign, to obtain

$$-a_{11} a_{23} a_{32} - a_{12} a_{21} a_{33} - a_{13} a_{22} a_{31}$$

The sum of all six terms exactly equals the earlier formula for  $|\mathbf{A}|$ .

Note that this method, known as **Sarrus's rule**, **does not generalize** to determinants of order higher than 3.

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# The Determinant Mapping

Let  $\mathcal{D}_n$  denote the domain  $\mathbb{R}^{n \times n}$  of  $n \times n$  matrices.

## Definition

For all  $n \in \mathbb{N}$ , the **determinant mapping**

$$\mathcal{D}_n \ni \mathbf{A} \mapsto |\mathbf{A}| := \sum_{\pi \in \Pi_n} \text{sgn}(\pi) \prod_{i=1}^n a_{i\pi(i)} \in \mathbb{R}$$

specifies the determinant  $|\mathbf{A}|$  of each  $n \times n$  matrix  $\mathbf{A}$  as a function of its  $n$  row vectors  $(\mathbf{a}_i^\top)_{i=1}^n$ . □

Here the multiplier  $\text{sgn}(\pi)$  attached to each product of  $n$  terms can be regarded as the **sign adjustment** associated with the permutation  $\pi \in \Pi_n$ .

## Row Mappings

For a general natural number  $n \in \mathbb{N}$ , consider any **row mapping**

$$\mathcal{D}_n \ni \mathbf{A} \mapsto D(\mathbf{A}) = D\left(\langle \mathbf{a}_i^\top \rangle_{i=1}^n\right) \in \mathbb{R}$$

defined on the domain  $\mathcal{D}_n$  of  $n \times n$  matrices  $\mathbf{A}$  with row vectors  $\langle \mathbf{a}_i^\top \rangle_{i=1}^n$ .

**Notation:** For each fixed  $r \in \mathbb{N}_n$ , let  $D(\mathbf{A}/\mathbf{b}_r^\top)$  denote the new value  $D(\mathbf{a}_1^\top, \dots, \mathbf{a}_{r-1}^\top, \mathbf{b}_r^\top, \mathbf{a}_{r+1}^\top, \dots, \mathbf{a}_n^\top)$  of the row mapping  $D$  after the  $r$ th row  $\mathbf{a}_r^\top$  of the matrix  $\mathbf{A}$  has been replaced by the new row vector  $\mathbf{b}_r^\top \in \mathbb{R}^n$ .

# Row Multilinearity

## Definition

The function  $\mathcal{D}_n \ni \mathbf{A} \mapsto D(\mathbf{A})$  of the  $n$  rows  $\langle \mathbf{a}_i^\top \rangle_{i=1}^n$  of  $\mathbf{A}$  is **(row) multilinear** just in case,  
for each row number  $i \in \{1, 2, \dots, n\}$ ,  
each pair  $\mathbf{b}_i^\top, \mathbf{c}_i^\top \in \mathbb{R}^n$  of new versions of row  $i$ ,  
and each pair of scalars  $\lambda, \mu \in \mathbb{R}$ , one has

$$D(\mathbf{A}_{-i}/\lambda\mathbf{b}_i^\top + \mu\mathbf{c}_i^\top) = \lambda D(\mathbf{A}_{-i}/\mathbf{b}_i^\top) + \mu D(\mathbf{A}_{-i}/\mathbf{c}_i^\top) \quad \square$$

Formally, the mapping  $\mathbb{R}^n \ni \mathbf{a}_i^\top \mapsto D(\mathbf{A}_{-i}/\mathbf{a}_i^\top) \in \mathbb{R}$   
is required to be linear, for fixed each row  $i \in \mathbb{N}_n$ .

That is,  $D$  is a linear function of the  $i$ th row vector  $\mathbf{a}_i^\top$  on its own,  
when all the other rows  $\mathbf{a}_h^\top$  ( $h \neq i$ ) are fixed.

# Determinants are Row Multilinear

## Theorem

For all  $n \in \mathbb{N}$ , the determinant mapping

$$\mathcal{D}_n \ni \mathbf{A} \mapsto |\mathbf{A}| := \sum_{\pi \in \Pi_n} \operatorname{sgn}(\pi) \prod_{i=1}^n a_{i\pi(i)} \in \mathbb{R}$$

is a row multilinear function of its  $n$  row vectors  $(\mathbf{a}_i^\top)_{i=1}^n$ .

## Proof.

For each fixed row  $r \in \mathbb{N}$ , we have

$$\begin{aligned} & \det(\mathbf{A}_{-i} / \lambda \mathbf{b}_r^\top + \mu \mathbf{c}_r^\top) \\ &= \sum_{\pi \in \Pi_n} \operatorname{sgn}(\pi) (\lambda b_{r\pi(r)} + \mu c_{r\pi(r)}) \prod_{i \neq r} a_{i\pi(i)} \\ &= \sum_{\pi \in \Pi_n} \operatorname{sgn}(\pi) \left[ \lambda b_{r\pi(r)} \prod_{i \neq r} a_{i\pi(i)} + \mu c_{r\pi(r)} \prod_{i \neq r} a_{i\pi(i)} \right] \\ &= \lambda \det(\mathbf{A}_{-i} / \mathbf{b}_r^\top) + \mu \det(\mathbf{A}_{-i} / \mathbf{c}_r^\top) \end{aligned}$$

as required for multilinearity. □

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# Permutation Matrices: Definition

## Definition

Given any permutation  $\pi \in \Pi_n$  on  $\{1, 2, \dots, n\}$ ,

define  $\mathbf{P}^\pi$  as the  $n \times n$  **permutation matrix**

whose elements satisfy  $p_{\pi(i),j}^\pi = \delta_{i,j}$  or equivalently  $p_{i,j}^\pi = \delta_{\pi^{-1}(i),j}$ .

That is, the rows of the identity matrix  $\mathbf{I}_n$  are permuted

so that for each  $i = 1, 2, \dots, n$ ,

its  $i$ th row vector is moved to become row  $\pi(i)$  of  $\mathbf{P}^\pi$ . □

## Lemma

For each permutation matrix  $\mathbf{P}^\pi$  one has  $(\mathbf{P}^\pi)^\top = \mathbf{P}^{\pi^{-1}}$ .

## Proof.

Because  $\pi$  is a permutation,  $i = \pi(j) \iff j = \pi^{-1}(i)$ .

Then the definitions imply that for all  $(i, j) \in \mathbb{N}_n^2$  one has

$$(\mathbf{P}^\pi)_{i,j}^\top = p_{j,i}^\pi = \delta_{\pi(j),i} = \delta_{\pi^{-1}(i),j} = p^{\pi^{-1}}(i,j) \quad \square$$

## Permutation Matrices: Examples

### Example

There are two  $2 \times 2$  permutation matrices, which are given by:

$$\mathbf{P}^{12} = \mathbf{I}_2; \quad \mathbf{P}^{21} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}.$$

Their signs are respectively  $+1$  and  $-1$ .

There are  $3! = 6$  permutation matrices in 3 dimensions given by:

$$\mathbf{P}^{123} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}; \quad \mathbf{P}^{132} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}; \quad \mathbf{P}^{213} = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix};$$

$$\mathbf{P}^{231} = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix}; \quad \mathbf{P}^{312} = \begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}; \quad \mathbf{P}^{321} = \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}.$$

Their signs are respectively  $+1$ ,  $-1$ ,  $-1$ ,  $+1$ ,  $+1$  and  $-1$ .

# Multiplying a Matrix by a Permutation Matrix

## Lemma

Given any  $n \times n$  matrix  $\mathbf{A}$ , for each permutation  $\pi \in \Pi_n$  the corresponding permutation matrix  $\mathbf{P}^\pi$  satisfies

$$(\mathbf{P}^\pi \mathbf{A})_{\pi(i),j} = a_{ij} = (\mathbf{A} \mathbf{P}^\pi)_{i,\pi(j)}$$

## Proof.

For each pair  $(i, j) \in \mathbb{N}_n^2$ , one has

$$(\mathbf{P}^\pi \mathbf{A})_{\pi(i),j} = \sum_{k=1}^n p_{\pi(i),k}^\pi a_{kj} = \sum_{k=1}^n \delta_{ik} a_{kj} = a_{ij}$$

and also

$$(\mathbf{A} \mathbf{P}^\pi)_{i,\pi(j)} = \sum_{k=1}^n a_{ik} p_{k,\pi(j)}^\pi = \sum_{k=1}^n a_{ik} \delta_{kj} = a_{ij} \quad \square$$

So  $\left\{ \begin{array}{l} \text{premultiplying} \\ \text{postmultiplying} \end{array} \right\} \mathbf{A}$  by  $\mathbf{P}^\pi$  applies  $\pi$  to  $\mathbf{A}$ 's  $\left\{ \begin{array}{l} \text{rows} \\ \text{columns} \end{array} \right\}$ .



# Multiplying Permutation Matrices

## Theorem

Given the composition  $\pi \circ \rho$  of two permutations  $\pi, \rho \in \Pi_n$ , the associated permutation matrices satisfy  $\mathbf{P}^\pi \mathbf{P}^\rho = \mathbf{P}^{\pi \circ \rho}$ .

## Proof.

For each pair  $(i, j) \in \mathbb{N}_n^2$ , one has

$$\begin{aligned}(\mathbf{P}^\pi \mathbf{P}^\rho)_{ij} &= \sum_{k=1}^n p_{ik}^\pi p_{kj}^\rho = \sum_{k=1}^n \delta_{\pi^{-1}(i), k} \delta_{\rho^{-1}(k), j} \\ &= \sum_{k=1}^n \delta_{(\rho^{-1} \circ \pi^{-1})(i), \rho^{-1}(k)} \delta_{\rho^{-1}(k), j} \\ &= \sum_{\ell=1}^n \delta_{(\pi \circ \rho)^{-1}(i), \ell} \delta_{\ell, j} = \delta_{(\pi \circ \rho)^{-1}(i), j} = p_{ij}^{\pi \circ \rho} \quad \square\end{aligned}$$

## Corollary

If  $\pi = \pi^1 \circ \pi^2 \circ \dots \circ \pi^q$ , then  $\mathbf{P}^\pi = \mathbf{P}^{\pi^1} \mathbf{P}^{\pi^2} \dots \mathbf{P}^{\pi^q}$ .

## Proof.

By induction on  $q$ , using the result of the Theorem. □

# Any Permutation Matrix Is Orthogonal

## Proposition

Any permutation matrix  $\mathbf{P}^\pi$  satisfies  $\mathbf{P}^\pi (\mathbf{P}^\pi)^\top = (\mathbf{P}^\pi)^\top \mathbf{P}^\pi = \mathbf{I}_n$ , so is orthogonal.

## Proof.

For each pair  $(i, j) \in \mathbb{N}_n^2$ , one has

$$\begin{aligned} [\mathbf{P}^\pi (\mathbf{P}^\pi)^\top]_{ij} &= \sum_{k=1}^n p_{ik}^\pi p_{jk}^\pi = \sum_{k=1}^n \delta_{\pi^{-1}(i), k} \delta_{\pi^{-1}(j), k} \\ &= \delta_{\pi^{-1}(i), \pi^{-1}(j)} = \delta_{ij} \end{aligned}$$

and also

$$\begin{aligned} [(\mathbf{P}^\pi)^\top \mathbf{P}^\pi]_{ij} &= \sum_{k=1}^n p_{ki}^\pi p_{kj}^\pi = \sum_{k=1}^n \delta_{\pi^{-1}(k), i} \delta_{\pi^{-1}(k), j} \\ &= \sum_{\ell=1}^n \delta_{\ell, i} \delta_{\ell, j} = \delta_{ij} \quad \square \end{aligned}$$

# Transposition Matrices

A special case of a permutation matrix is a **transposition**  $\mathbf{T}_{rs}$  of rows  $r$  and  $s$ .

As the matrix  $\mathbf{I}$  with rows  $r$  and  $s$  transposed, it satisfies

$$(\mathbf{T}_{rs})_{ij} = \begin{cases} \delta_{ij} & \text{if } i \notin \{r, s\} \\ \delta_{sj} & \text{if } i = r \\ \delta_{rj} & \text{if } i = s \end{cases}$$

## Exercise

Let  $\mathbf{A}$  be any  $n \times n$  matrix. Prove that:

- 1) any transposition matrix  $\mathbf{T}_{rs}$  is symmetric and orthogonal;
- 2)  $\mathbf{T}_{rs} = \mathbf{T}_{sr}$ ;    3)  $\mathbf{T}_{rs}\mathbf{T}_{sr} = \mathbf{T}_{sr}\mathbf{T}_{rs} = \mathbf{I}$ ;
- 4)  $\mathbf{T}_{rs}\mathbf{A}$  is  $\mathbf{A}$  with rows  $r$  and  $s$  interchanged;
- 5)  $\mathbf{A}\mathbf{T}_{rs}$  is  $\mathbf{A}$  with columns  $r$  and  $s$  interchanged.

# Determinants with Permuted Rows: Theorem

## Theorem

Given any  $n \times n$  matrix  $\mathbf{A}$  and any permutation  $\pi \in \mathbb{N}_n$ , one has  $|\mathbf{P}^\pi \mathbf{A}| = |\mathbf{A} \mathbf{P}^\pi| = \text{sgn}(\pi) |\mathbf{A}|$ .

# Determinants with Permuted Rows: Proof

Proof.

The expansion formula for determinants gives

$$|\mathbf{P}^\pi \mathbf{A}| = \sum_{\rho \in \Pi_n} \operatorname{sgn}(\rho) \prod_{i=1}^n (\mathbf{P}^\pi \mathbf{A})_{i,\rho(i)}$$

But for each  $i \in \mathbb{N}_n$ ,  $\rho \in \Pi_n$ , one has  $(\mathbf{P}^\pi \mathbf{A})_{i,\rho(i)} = a_{\pi^{-1}(i),\rho(i)}$ , so

$$\begin{aligned} |\mathbf{P}^\pi \mathbf{A}| &= \sum_{\rho \in \Pi_n} \operatorname{sgn}(\rho) \prod_{i=1}^n a_{\pi^{-1}(i),\rho(i)} \\ &= [1/\operatorname{sgn}(\pi)] \sum_{\pi \circ \rho \in \Pi_n} \operatorname{sgn}(\pi \circ \rho) \prod_{i=1}^n a_{i,(\pi \circ \rho)(i)} \\ &= \operatorname{sgn}(\pi) \sum_{\sigma \in \Pi_n} \operatorname{sgn}(\sigma) \prod_{i=1}^n a_{i,\sigma(i)} = \operatorname{sgn}(\pi) |\mathbf{A}| \end{aligned}$$

because  $\operatorname{sgn}(\pi \circ \rho) = \operatorname{sgn}(\pi) \operatorname{sgn}(\rho)$  and  $1/\operatorname{sgn}(\pi) = \operatorname{sgn}(\pi)$ , whereas there is an obvious bijection  $\Pi_n \ni \rho \leftrightarrow \pi \circ \rho = \sigma \in \Pi_n$  on the set of permutations  $\Pi_n$ .

The proof that  $|\mathbf{A}\mathbf{P}^\pi| = \operatorname{sgn}(\pi) |\mathbf{A}|$  is sufficiently similar to be left as an exercise. □

# The Alternation Rule for Determinants

## Corollary

Given any  $n \times n$  matrix  $\mathbf{A}$

and any transposition  $\tau_{rs}$  with associated transposition matrix  $\mathbf{T}_{rs}$ ,  
one has  $|\mathbf{T}_{rs}\mathbf{A}| = |\mathbf{A}\mathbf{T}_{rs}| = -|\mathbf{A}|$ .

## Proof.

Apply the previous theorem in the special case  
when  $\pi = \tau_{rs}$  and so  $\mathbf{P}^\pi = \mathbf{T}_{rs}$ .

Then, because  $\text{sgn}(\pi) = \text{sgn}(\tau_{rs}) = -1$ ,

the equality  $|\mathbf{P}^\pi\mathbf{A}| = \text{sgn}(\pi)|\mathbf{A}|$  implies that  $|\mathbf{T}_{rs}\mathbf{A}| = -|\mathbf{A}|$ . □

We have shown that, for any  $n \times n$  matrix  $\mathbf{A}$ , given any:

1. permutation  $\pi \in \mathbb{N}_n$ , one has  $|\mathbf{P}^\pi\mathbf{A}| = |\mathbf{A}\mathbf{P}^\pi| = \text{sgn}(\pi)|\mathbf{A}|$ ;
2. transposition  $\tau_{rs}$ , one has  $|\mathbf{T}_{rs}\mathbf{A}| = |\mathbf{A}\mathbf{T}_{rs}| = -|\mathbf{A}|$ .

## Sign Adjusted Transpositions

We define the **sign adjusted** transposition matrix  $\hat{\mathbf{T}}_{rs}$  as either one of the two matrices that:

- (i) swaps rows or columns  $r$  and  $s$ ;
- (ii) then multiplies one, but only one, of the two swapped rows or columns by  $-1$ .

As the matrix  $\mathbf{I}$  with rows  $r$  and  $s$  transposed, and then one sign changed, it satisfies

$$(\mathbf{T}_{rs})_{ij} = \begin{cases} \delta_{ij} & \text{if } i \notin \{r, s\} \\ \alpha_s \delta_{sj} & \text{if } i = r \\ \alpha_r \delta_{rj} & \text{if } i = s \end{cases}$$

where  $\alpha_r, \alpha_s \in \{-1, +1\}$  with  $\alpha_r = -\alpha_s$ .

It evidently satisfies  $|\hat{\mathbf{T}}_{rs}\mathbf{A}| = |\mathbf{A}\hat{\mathbf{T}}_{rs}| = |\mathbf{A}|$ .

## Sign Adjusted Permutations

Given any permutation matrix  $\mathbf{P}$ ,  
there is a unique permutation  $\pi$  such that  $\mathbf{P} = \mathbf{P}^\pi$ .

Suppose that  $\pi = \tau_{r_1 s_1} \circ \cdots \circ \tau_{r_\ell s_\ell}$  is any one of the several ways  
in which the permutation  $\pi$  can be decomposed  
into a composition of transpositions.

Then  $\mathbf{P} = \prod_{k=1}^{\ell} \mathbf{T}_{r_k s_k}$  and  $|\mathbf{PA}| = (-1)^\ell |\mathbf{A}|$  for any  $\mathbf{A}$ .

### Definition

Say that  $\hat{\mathbf{P}}$  is a **sign adjusted** version of  $\mathbf{P} = \mathbf{P}^\pi$   
just in case it can be expressed as the product  $\hat{\mathbf{P}} = \prod_{k=1}^{\ell} \hat{\mathbf{T}}_{r_k s_k}$   
of sign adjusted transpositions satisfying  $\mathbf{P} = \prod_{k=1}^{\ell} \mathbf{T}_{r_k s_k}$ .

Then it is easy to prove by induction on  $\ell$   
that for every  $n \times n$  matrix  $\mathbf{A}$  one has  $|\hat{\mathbf{P}}\mathbf{A}| = |\mathbf{A}\hat{\mathbf{P}}| = |\mathbf{A}|$ .

Recall that all the elements of a permutation matrix  $\mathbf{P}$  are 0 or 1.

A sign adjustment of  $\mathbf{P}$  involves changing some of the 1 elements  
into  $-1$  elements, while leaving all the 0 elements unchanged.



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# Triangular Matrices: Definition

## Definition

A square matrix is **upper** (resp. **lower**) **triangular** if all its non-zero off diagonal elements are above and to the right (resp. below and to the left) of the diagonal — i.e., in the upper (resp. lower) triangle bounded by the principal diagonal.

- ▶ The elements of an upper triangular matrix **U** satisfy  $(\mathbf{U})_{ij} = 0$  whenever  $i > j$ .
- ▶ The elements of a lower triangular matrix **L** satisfy  $(\mathbf{L})_{ij} = 0$  whenever  $i < j$ .

# Products of Upper Triangular Matrices

## Theorem

The product  $\mathbf{W} = \mathbf{UV}$  of any two upper triangular matrices  $\mathbf{U}, \mathbf{V}$  is upper triangular,

with diagonal elements  $w_{ii} = u_{ii}v_{ii}$  ( $i = 1, \dots, n$ ) equal to the product of the corresponding diagonal elements of  $\mathbf{U}, \mathbf{V}$ .

## Proof.

Given any two upper triangular  $n \times n$  matrices  $\mathbf{U}$  and  $\mathbf{V}$ , one has  $u_{ik}v_{kj} = 0$  unless both  $i \leq k$  and  $k \leq j$ .

So the elements  $(w_{ij})^{n \times n}$  of their product  $\mathbf{W} = \mathbf{UV}$  satisfy

$$w_{ij} = \begin{cases} \sum_{k=i}^j u_{ik}v_{kj} & \text{if } i \leq j \\ 0 & \text{if } i > j \end{cases}$$

Hence  $\mathbf{W} = \mathbf{UV}$  is upper triangular.

Finally, when  $j = i$  the above sum collapses to just one term, and  $w_{ii} = u_{ii}v_{ii}$  for  $i = 1, \dots, n$ . □

# Triangular Matrices: Exercises

## Exercise

*Prove that the transpose:*

1.  $\mathbf{U}^\top$  of any upper triangular matrix  $\mathbf{U}$  is lower triangular;
2.  $\mathbf{L}^\top$  of any lower triangular matrix  $\mathbf{L}$  is upper triangular.

## Exercise

*Consider the matrix  $\mathbf{E}_{r+\alpha q}$  that represents the elementary row operation of adding a multiple of  $\alpha$  times row  $q$  to row  $r$ , with  $r \neq q$ .*

*Under what conditions is  $\mathbf{E}_{r+\alpha q}$  (i) upper triangular? (ii) lower triangular?*

**Hint:** Apply the row operation to the identity matrix  $\mathbf{I}$ .

**Answer:** (i) iff  $q < r$ ; (ii) iff  $q > r$ .

# Products of Lower Triangular Matrices

## Theorem

*The product of any two lower triangular matrices is lower triangular.*

## Proof.

Given any two lower triangular matrices  $\mathbf{L}$ ,  $\mathbf{M}$ , taking transposes shows that  $(\mathbf{LM})^\top = \mathbf{M}^\top \mathbf{L}^\top = \mathbf{U}$ , where the product  $\mathbf{U}$  is upper triangular, as the product of upper triangular matrices.

Hence  $\mathbf{LM} = \mathbf{U}^\top$  is lower triangular, as the transpose of an upper triangular matrix. □

# Determinants of Triangular Matrices

## Theorem

*The determinant of any  $n \times n$  upper triangular matrix  $\mathbf{U}$  equals the product of all the elements on its principal diagonal.*

## Proof.

Recall the expansion formula  $|\mathbf{U}| = \sum_{\pi \in \Pi} \text{sgn}(\pi) \prod_{i=1}^n u_{i\pi(i)}$  where  $\Pi$  denotes the set of permutations on  $\{1, 2, \dots, n\}$ .

Because  $\mathbf{U}$  is upper triangular, one has  $u_{i\pi(i)} = 0$  unless  $i \leq \pi(i)$ .

So  $\prod_{i=1}^n u_{i\pi(i)} = 0$  unless  $i \leq \pi(i)$  for all  $i = 1, 2, \dots, n$ .

But the identity  $\iota$  is the only permutation  $\pi \in \Pi$  that satisfies  $i \leq \pi(i)$  for all  $i \in \mathbb{N}_n$ .

Because  $\text{sgn}(\iota) = +1$ , the expansion reduces to the single term

$$|\mathbf{U}| = \text{sgn}(\iota) \prod_{i=1}^n u_{i\iota(i)} = \prod_{i=1}^n u_{ii}$$

This is the product of the  $n$  diagonal elements, as claimed. □

# Invertible Triangular Matrices

Similarly  $|\mathbf{L}| = \prod_{i=1}^n \ell_{ii}$  for any lower triangular matrix  $\mathbf{L}$ .

Evidently:

## Corollary

*A triangular matrix (upper or lower) is invertible if and only if no element on its principal diagonal is 0.*