

# Lecture Notes: Matrix Algebra

## Part B: Introduction to Matrices

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# Outline

## Matrices: Introduction

Matrices and Their Transposes

Matrix Multiplication: Definition

## Special Matrices

Square, Symmetric, and Diagonal Matrices

The Identity Matrix

The Inverse Matrix

Partitioned Matrices

Block Diagonal Matrices

## Permutations and Their Signs

Permutations

Transpositions

Adjacency Transpositions

The Inversions and Sign of a Permutation

The Product Rule

## Matrices as Rectangular Arrays

An  $m \times n$  **matrix**  $\mathbf{A} = (a_{ij})_{m \times n}$  is a (rectangular) array, such as

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} = ((a_{ij})_{i=1}^m)_{j=1}^n = ((a_{ij})_{j=1}^n)_{i=1}^m$$

Note that in  $a_{ij}$ , we write the **row** number  $i$  **before** the **column** number  $j$ .

An  $m \times 1$  matrix is a **column vector** with  $m$  rows and 1 column.

A  $1 \times n$  matrix is a **row vector** with 1 row and  $n$  columns.

The  $m \times n$  **matrix**  $\mathbf{A}$  consists of:

$n$  **columns** in the form of  $m$ -vectors

$$\mathbf{a}_j = (a_{ij})_{i=1}^m \in \mathbb{R}^m \text{ for } j = 1, 2, \dots, n;$$

$m$  **rows** in the form of  $n$ -vectors

$$\mathbf{a}_i^\top = (a_{ij})_{j=1}^n \in \mathbb{R}^n \text{ for } i = 1, 2, \dots, m.$$

# The Transpose of a Matrix

The **transpose**  $\mathbf{A}^\top$  of the  $m \times n$  matrix  $\mathbf{A} = (a_{ij})_{m \times n}$  is defined as the  $n \times m$  matrix

$$\mathbf{A}^\top = (a_{ij}^\top)_{n \times m} = (a_{ji})_{n \times m} = \begin{pmatrix} a_{11} & a_{21} & \dots & a_{n1} \\ a_{12} & a_{22} & \dots & a_{n2} \\ \vdots & \vdots & & \vdots \\ a_{1m} & a_{2m} & \dots & a_{nm} \end{pmatrix}$$

Thus the transposed matrix  $\mathbf{A}^\top$  results from transforming each column  $m$ -vector  $\mathbf{a}_j = (a_{ij})_{i=1}^m$  ( $j = 1, 2, \dots, n$ ) of  $\mathbf{A}$  into the corresponding row  $m$ -vector  $\mathbf{a}_j^\top = (a_{ij}^\top)_{i=1}^m$  of  $\mathbf{A}^\top$ .

Equivalently, for each  $i = 1, 2, \dots, m$ , the  $i$ th row  $n$ -vector  $\mathbf{a}_i^\top = (a_{ij})_{j=1}^n$  of  $\mathbf{A}$  is transformed into the  $i$ th column  $n$ -vector  $\mathbf{a}_i = (a_{ji})_{j=1}^n$  of  $\mathbf{A}^\top$ .

Either way, one has  $a_{ij}^\top = a_{ji}$  for all relevant pairs  $i, j$ .

# Rows Before Columns

**VERY Important Rule:** Rows **before** columns!

This order really matters.

Reversing it gives a transposed matrix.

## Exercise

*Verify that the double transpose of any  $m \times n$  matrix  $\mathbf{A}$  satisfies  $(\mathbf{A}^\top)^\top = \mathbf{A}$*

*— i.e., transposing a matrix twice recovers the original matrix.*

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Transpositions

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# Multiplying a Matrix by a Scalar

A **scalar**, usually denoted by a Greek letter, is simply a member  $\alpha \in \mathbb{F}$  of the algebraic field  $\mathbb{F}$  over which the vector space is defined.

So when  $\mathbb{F} = \mathbb{R}$ , a scalar is a real number  $\alpha \in \mathbb{R}$ .

The **product** of any  $m \times n$  matrix  $\mathbf{A} = (a_{ij})_{m \times n}$  and any scalar  $\alpha \in \mathbb{R}$  is the new  $m \times n$  matrix denoted by  $\alpha\mathbf{A} = (\alpha a_{ij})_{m \times n}$ , each of whose elements  $\alpha a_{ij}$  results from multiplying the corresponding element  $a_{ij}$  of  $\mathbf{A}$  by  $\alpha$ .

# Matrix Multiplication

The **matrix product** of two matrices **A** and **B** is defined (whenever possible) as the matrix  $\mathbf{C} = \mathbf{AB} = (c_{ij})_{m \times n}$  whose element  $c_{ij}$  in row  $i$  and column  $j$  is the inner product  $c_{ij} = \mathbf{a}_i^\top \mathbf{b}_j$  of:

- ▶ the  $i$ th **row** vector  $\mathbf{a}_i^\top$  of the first matrix **A**;
- ▶ the  $j$ th **column** vector  $\mathbf{b}_j$  of the second matrix **B**.

$$\begin{pmatrix} a_{11} & \dots & a_{1j} & \dots & a_{1n} \\ \vdots & & \vdots & & \vdots \\ a_{i1} & \dots & a_{ij} & \dots & a_{in} \\ \vdots & & \vdots & & \vdots \\ a_{m1} & \dots & a_{mj} & \dots & a_{mn} \end{pmatrix} \begin{pmatrix} b_{11} & \dots & b_{1j} & \dots & b_{1p} \\ \vdots & & \vdots & & \vdots \\ b_{j1} & \dots & b_{jj} & \dots & b_{jp} \\ \vdots & & \vdots & & \vdots \\ b_{m1} & \dots & b_{mj} & \dots & b_{mp} \end{pmatrix} = \begin{pmatrix} c_{11} & \dots & c_{1j} & \dots & c_{1p} \\ \vdots & & \vdots & & \vdots \\ c_{i1} & \dots & c_{ij} & \dots & c_{ip} \\ \vdots & & \vdots & & \vdots \\ c_{m1} & \dots & c_{mj} & \dots & c_{mp} \end{pmatrix}$$

$\mathbf{a}_i^\top \quad \cdot \quad \mathbf{b}_j \quad = \quad c_{ij}$

Again: rows **before** columns!

# Compatibility for Matrix Multiplication, I

Note that the resulting matrix product  $\mathbf{C}$  must have:

- ▶ as many rows as the first matrix  $\mathbf{A}$ ;
- ▶ as many columns as the second matrix  $\mathbf{B}$ .

Yet again: rows **before** columns!

## Compatibility for Matrix Multiplication, II

**Question:** when is this definition  
of the matrix product  $\mathbf{C} = \mathbf{AB}$  possible?

**Answer:** if and only if  $\mathbf{A}$  has as many columns as  $\mathbf{B}$  has rows.

This condition ensures that every inner product  $\mathbf{a}_i^\top \mathbf{b}_j$  is defined, which is true iff (if and only if) every row of  $\mathbf{A}$  has exactly the same number of elements as every column of  $\mathbf{B}$ .

In this case, the two matrices  $\mathbf{A}$  and  $\mathbf{B}$  are **compatible for multiplication**.

Specifically, if  $\mathbf{A}$  is  $m \times \ell$  for some  $m$ , then  $\mathbf{B}$  must be  $\ell \times n$  for some  $n$ .

Then the product  $\mathbf{C} = \mathbf{AB}$  is  $m \times n$ , with elements  $c_{ij} = \mathbf{a}_i^\top \mathbf{b}_j = \sum_{k=1}^{\ell} a_{ik} b_{kj}$  for  $i = 1, 2, \dots, m$  and  $j = 1, 2, \dots, n$ .

# Laws of Matrix Multiplication

## Exercise

Verify that, whenever the relevant matrix products are defined, the following *laws of matrix multiplication* hold:

*associative law for matrices:*  $\mathbf{A}(\mathbf{BC}) = (\mathbf{AB})\mathbf{C}$ ;

*distributive:*  $\mathbf{A}(\mathbf{B} + \mathbf{C}) = \mathbf{AB} + \mathbf{AC}$  and  $(\mathbf{A} + \mathbf{B})\mathbf{C} = \mathbf{AC} + \mathbf{BC}$ ;

*transpose:*  $(\mathbf{AB})^\top = \mathbf{B}^\top \mathbf{A}^\top$ .

*associative law for scalars:*  $\alpha(\mathbf{AB}) = (\alpha\mathbf{A})\mathbf{B} = \mathbf{A}(\alpha\mathbf{B})$  (all  $\alpha \in \mathbb{R}$ ).

## Exercise

Let  $\mathbf{X}$  be any  $m \times n$  matrix, and  $\mathbf{z}$  any column  $n$ -vector.

1. Show that the matrix product  $\mathbf{z}^\top \mathbf{X}^\top \mathbf{X} \mathbf{z}$  is well-defined, and that its value is a scalar.
2. By putting  $\mathbf{w} = \mathbf{X} \mathbf{z}$  in the previous exercise regarding the sign of the quadratic form  $\mathbf{w}^\top \mathbf{w}$ , what can you conclude about the value of the scalar  $\mathbf{z}^\top \mathbf{X}^\top \mathbf{X} \mathbf{z}$ ?

# Exercise for Econometricians I

## Exercise

An econometrician has access to data in the form of the real-valued time series:

- ▶  $y_t$  ( $t = 1, 2, \dots, T$ ) of one *endogenous* variable;
- ▶  $x_{ti}$  ( $t = 1, 2, \dots, T$  and  $i = 1, 2, \dots, k$ ) of  $k$  different *exogenous* variables  
— sometimes called *explanatory* variables or *regressors*.

The data is to be fitted to the *linear regression model*

$$y_t = \sum_{i=1}^k b_i x_{ti} + e_t$$

whose scalar constants  $b_i$  ( $i = 1, 2, \dots, k$ ) are unknown *regression coefficients*, and each scalar  $e_t$  is the *error term* or *residual*.

## Exercise for Econometricians II

1. Discuss how the regression model with  $y_t = \sum_{i=1}^k b_i x_{ti} + e_t$  can be written in the form  $\mathbf{y} = \mathbf{X}\mathbf{b} + \mathbf{e}$  for suitable column vectors  $\mathbf{y}$ ,  $\mathbf{b}$ ,  $\mathbf{e}$ .
2. What are the dimensions of these vectors, and of the exogenous data matrix  $\mathbf{X}$ ?
3. Why do you think econometricians use this matrix equation, rather than the alternative  $\mathbf{y} = \mathbf{b}\mathbf{X} + \mathbf{e}$ ?
4. How can the equation  $\mathbf{y} = \mathbf{X}\mathbf{b} + \mathbf{e}$  accommodate the constant term  $\alpha$  in the alternative equation  $y_t = \alpha + \sum_{i=1}^k b_i x_{ti} + e_t$ ?

## Matrix Multiplication Does Not Commute I

The two matrices **A** and **B** **commute** just in case **AB = BA**.

Note that typical pairs of matrices **DO NOT** commute, meaning that **AB  $\neq$  BA** — i.e., the order of multiplication matters.

Indeed, suppose that **A** is  $\ell \times m$  and **B** is  $m \times n$ , as is needed for **AB** to be defined as an  $\ell \times n$  matrix.

Then the reverse product **BA** is **undefined** except in the special case when  $n = \ell$ .

Hence, for both **AB** and **BA** to be defined, where **B** is  $m \times n$ , the matrix **A** **must** be  $n \times m$ .

But then **AB** is  $n \times n$ , whereas **BA** is  $m \times m$ .

Evidently **AB  $\neq$  BA** unless  $m = n$ .

Then all four matrices **A**, **B**, **AB** and **BA** are  $m \times m = n \times n$ .

To summarize, we must be in the special case where **A** and **B** are two **square** matrices of the **same** dimension.

## Matrix Multiplication Does Not Commute II

Even if both **A** and **B** are  $n \times n$  matrices, implying that both **AB** and **BA** are also  $n \times n$ , one can still have **AB**  $\neq$  **BA**.

### Example

Here is a  $2 \times 2$  example:

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

### Exercise

*For matrix multiplication, explain why there are two different versions of the distributive law — namely*

$$\mathbf{A}(\mathbf{B} + \mathbf{C}) = \mathbf{AB} + \mathbf{AC} \text{ and } (\mathbf{A} + \mathbf{B})\mathbf{C} = \mathbf{AC} + \mathbf{BC}$$

# More Warnings Regarding Matrix Multiplication

## Exercise

Let  $\mathbf{A}$ ,  $\mathbf{B}$ ,  $\mathbf{C}$  denote three general matrices.

Give examples showing that:

1. The matrix  $\mathbf{AB}$  might be defined, even if  $\mathbf{BA}$  is not.
2. One can have  $\mathbf{AB} = \mathbf{0}$  even though  $\mathbf{A} \neq \mathbf{0}$  and  $\mathbf{B} \neq \mathbf{0}$ .
3. If  $\mathbf{AB} = \mathbf{AC}$  and  $\mathbf{A} \neq \mathbf{0}$ , it does not follow that  $\mathbf{B} = \mathbf{C}$ .

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## Special Matrices

Square, Symmetric, and Diagonal Matrices

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Permutations

Transpositions

Adjacency Transpositions

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The Product Rule

# 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

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

The extra dots indicate omitted elements 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

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

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  $\mathbf{diag} \mathbf{d} = (a_{ij})_{n \times n}$  where each  $a_{ij} = \delta_{ij} d_i = \delta_{ij} d_j$ , using Kronecker delta notation.

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

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

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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Transpositions

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The Product Rule

# 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  $\mathbb{N}_n \times \mathbb{N}_n \ni (i, j) \mapsto \delta_{ij} \in \{0, 1\}$ .

## 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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Square, Symmetric, and Diagonal Matrices

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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 of a square matrix exist?

**Answer** discussed later: if and only if its **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

## 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.*

# Rule for Inverting Transposes

## 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

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

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

This proves that  $(\mathbf{A}^{-1})^\top$  is both a left and a right inverse of  $\mathbf{A}^\top$ . □

# 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, one has  $\mathbf{x}_i \cdot \mathbf{x}_i = \|\mathbf{x}_i\|^2 = 1$  and so  $\|\mathbf{x}_i\| = 1$  for each  $i \in \mathbb{N}_k$  — i.e., all  $k$  elements of the set are vectors of unit length.

## Lemma

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

## Proof.

The result is immediate from the definitions of the norm in  $\mathbb{R}^n$ , as well as of an orthonormal set of vectors and of the Kronecker delta function. □

# 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.

(The answer will come later in the section on eigenvalues and eigenvectors.)

# Outline

## Matrices: Introduction

Matrices and Their Transposes

Matrix Multiplication: Definition

## Special Matrices

Square, Symmetric, and Diagonal Matrices

The Identity Matrix

The Inverse Matrix

**Partitioned Matrices**

Block Diagonal Matrices

## Permutations and Their Signs

Permutations

Transpositions

Adjacency Transpositions

The Inversions and Sign of a Permutation

The Product Rule

## 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)  $\mathbf{A}$  and  $\mathbf{E}$ ; (ii)  $\mathbf{B}$  and  $\mathbf{F}$ ; (iii)  $\mathbf{C}$  and  $\mathbf{G}$ ; (iv)  $\mathbf{D}$  and  $\mathbf{H}$ .

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

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

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

# Inverting Partitioned Matrices: Two Exercises

## Exercise

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

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?

# Outline

## Matrices: Introduction

Matrices and Their Transposes

Matrix Multiplication: Definition

## Special Matrices

Square, Symmetric, and Diagonal Matrices

The Identity Matrix

The Inverse Matrix

Partitioned Matrices

**Block Diagonal Matrices**

## Permutations and Their Signs

Permutations

Transpositions

Adjacency Transpositions

The Inversions and Sign of a Permutation

The Product Rule

# Block Diagonal Matrices: Definition

## Definition

A **block diagonal matrix** is a partitioned square matrix which is a diagonal  $k \times k$  square array of “blocks” in the form of  $n_i \times n_i$  square matrices  $\mathbf{A}_{n_i \times n_i}^{(i)}$ , for  $i \in \mathbb{N}_k$ .

The array can be written as

$$\begin{pmatrix} \mathbf{A}_{n_1 \times n_1}^{(1)} & \mathbf{0}_{n_1 \times n_2} & \cdots & \mathbf{0}_{n_1 \times n_k} \\ \mathbf{0}_{n_2 \times n_1} & \mathbf{A}_{n_2 \times n_2}^{(2)} & \cdots & \mathbf{0}_{n_2 \times n_k} \\ \vdots & \vdots & \ddots & \vdots \\ \mathbf{0}_{n_k \times n_1} & \mathbf{0}_{n_k \times n_2} & \cdots & \mathbf{A}_{n_k \times n_k}^{(k)} \end{pmatrix}$$

or, more succinctly, as  $\text{diag}(\mathbf{A}_{n_1 \times n_1}^{(1)}, \dots, \mathbf{A}_{n_k \times n_k}^{(k)})$ .

# Products of Block Diagonal Matrices

## Exercise

Suppose that the two block diagonal matrices

$$\text{diag}(\mathbf{A}_{m_1 \times m_1}^{(1)}, \dots, \mathbf{A}_{m_k \times m_k}^{(k)}) \quad \text{and} \quad \text{diag}(\mathbf{B}_{n_1 \times n_1}^{(1)}, \dots, \mathbf{B}_{n_\ell \times n_\ell}^{(\ell)})$$

have *compatible dimensions* in the sense that  $k = \ell$   
and  $m_i = n_i$  for all  $i \in \mathbb{N}_k = \mathbb{N}_\ell$ .

Verify that then the two matrix products

$$\begin{aligned} & \text{diag}(\mathbf{A}^{(1)}, \dots, \mathbf{A}^{(k)}) \text{diag}(\mathbf{B}^{(1)}, \dots, \mathbf{B}^{(k)}) \\ \text{and} \quad & \text{diag}(\mathbf{B}^{(1)}, \dots, \mathbf{B}^{(k)}) \text{diag}(\mathbf{A}^{(1)}, \dots, \mathbf{A}^{(k)}) \end{aligned}$$

both exist, and that they equal

$$\text{diag}(\mathbf{A}^{(1)}\mathbf{B}^{(1)}, \dots, \mathbf{A}^{(k)}\mathbf{B}^{(k)}) \quad \text{and} \quad \text{diag}(\mathbf{B}^{(1)}\mathbf{A}^{(1)}, \dots, \mathbf{B}^{(k)}\mathbf{A}^{(k)})$$

respectively.

# The Inverse of a Block Diagonal Matrix

## Exercise

Suppose the block diagonal matrix  $\mathbf{diag}(\mathbf{A}^{(1)}, \dots, \mathbf{A}^{(k)})$  has the property that each block  $\mathbf{A}_{n_i \times n_i}^{(i)}$  ( $i \in \mathbb{N}_k$ ) is invertible.

Show that then the block diagonal matrix is invertible, with inverse

$$\left[ \mathbf{diag}(\mathbf{A}^{(1)}, \dots, \mathbf{A}^{(k)}) \right]^{-1} = \mathbf{diag} \left( \left[ \mathbf{A}^{(1)} \right]^{-1}, \dots, \left[ \mathbf{A}^{(k)} \right]^{-1} \right)$$

# Outline

## Matrices: Introduction

Matrices and Their Transposes

Matrix Multiplication: Definition

## Special Matrices

Square, Symmetric, and Diagonal Matrices

The Identity Matrix

The Inverse Matrix

Partitioned Matrices

Block Diagonal Matrices

## Permutations and Their Signs

**Permutations**

Transpositions

Adjacency Transpositions

The Inversions and Sign of a Permutation

The Product Rule

# 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  $(\Pi_n, \circ, \iota)$  is an algebraic **group** — i.e., the operation  $\Pi_n \times \Pi_n \ni (\pi, \rho) \mapsto \pi \circ \rho \in \Pi_n$  is associative, with  $\iota$  as unit, and an inverse  $\pi^{-1}$  for every  $\pi \in \Pi_n$ .

# Outline

## Matrices: Introduction

Matrices and Their Transposes

Matrix Multiplication: Definition

## Special Matrices

Square, Symmetric, and Diagonal Matrices

The Identity Matrix

The Inverse Matrix

Partitioned Matrices

Block Diagonal Matrices

## Permutations and Their Signs

Permutations

**Transpositions**

Adjacency Transpositions

The Inversions and Sign of a Permutation

The Product Rule

# 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

Let  $(j_1, j_2, \dots, j_n) = (j_k)_{k \in \mathbb{N}_n} \in \mathbb{N}_n^n$  denote any list of  $n$  integers  $(j_k)_{k \in \mathbb{N}_n}$  in  $\mathbb{N}_n$ , or equivalently, any mapping  $\mathbb{N}_n \ni k \mapsto j_k \in \mathbb{N}_n$ .

Then any list  $(j_k)_{k \in \mathbb{N}_n}$  whose components in  $\mathbb{N}_n$  are all different corresponds to a unique permutation, denoted by  $\pi^{j_1 j_2 \dots j_n} \in \Pi_n$ , that satisfies  $\pi(k) = j_k$  for all  $k \in \mathbb{N}_n$ .

## Example

Two transpositions defined on a set containing more than two elements **may not commute**.

For example, one has

$$\tau_{12} \circ \tau_{23} = \tau_{12}(\pi^{132}) = \pi^{312} \quad \text{and} \quad \tau_{23} \circ \tau_{12} = \tau_{23}(\pi^{213}) = \pi^{231}$$

# 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 the Induction Step

For general  $n$ , let  $j := \pi^{-1}(n)$  be the element  $\pi$  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, the permutation  $\tilde{\pi}$  on  $\mathbb{N}_{n-1}$  is the product  $\tau^1 \circ \tau^2 \circ \dots \circ \tau^q$  of  $q \leq n - 2$  transpositions.

Hence, for all  $k \in \mathbb{N}_{n-1}$ , one has  $\tilde{\pi}(k) = (\pi \circ \tau_{jn})(k) = (\tau^1 \circ \tau^2 \circ \dots \circ \tau^q)(k)$ .

Also, for each  $p = 1, \dots, q$ , 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.

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

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

## Example Illustrating the Proof

### Example

We find a set of transpositions whose product is the particular permutation  $\pi^{4132} \in \Pi_4$ .

Using compressed notation, start with the transposition  $\tau_{14}$  which moves the 4 in 4132 to its right place in the end.

In fact  $\tau_{14}$  applied to  $\pi^{4132}$  is  $\pi^{2134}$ , which we abbreviate to 2134, with 3 and 4 in the right place.

Last, we use  $\tau_{12}$  to get 1234.

Reversing the order of these transpositions

gives us  $\pi^{4132} = \tau_{14} \circ \tau_{12}$

because  $(\tau_{14} \circ \tau_{12})(1234) = \tau_{14}(2134) = 4132$ .

# Outline

## Matrices: Introduction

Matrices and Their Transposes

Matrix Multiplication: Definition

## Special Matrices

Square, Symmetric, and Diagonal Matrices

The Identity Matrix

The Inverse Matrix

Partitioned Matrices

Block Diagonal Matrices

## Permutations and Their Signs

Permutations

Transpositions

**Adjacency Transpositions**

The Inversions and Sign of a Permutation

The Product Rule

# 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 the right 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  $\pi^{k+1} \nearrow \ell \circ \pi^\ell \searrow k$ .

Both are 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  $k$ th 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. □

## Example Illustrating the Proof

### Example

We illustrate how  $\tau_{2,5} = \pi^{4 \searrow 2} \circ \pi^{2 \nearrow 5}$ .

Indeed, applying  $\pi^{2 \nearrow 5}$  to 123456 gives successively

132456

134256

134526

Then, applying  $\pi^{4 \searrow 2}$  to 134526 gives successively

134526

135426

153426

The result 153426 is indeed  $\tau_{2,5}$  applied to 123456.

# Outline

## Matrices: Introduction

Matrices and Their Transposes

Matrix Multiplication: Definition

## Special Matrices

Square, Symmetric, and Diagonal Matrices

The Identity Matrix

The Inverse Matrix

Partitioned Matrices

Block Diagonal Matrices

## Permutations and Their Signs

Permutations

Transpositions

Adjacency Transpositions

**The Inversions and Sign of a Permutation**

The Product Rule

# The Inversions of a Permutation

## Definition

1. Let  $\mathbb{N}_{n,2}^< = \{(i, j) \in \mathbb{N}_n \times \mathbb{N}_n \mid i < j\}$  denote the set of all ordered **pair subsets** of  $\mathbb{N}_n$ .
2. Given any permutation  $\pi \in \Pi_n$ , the ordered pair  $(i, j) \in \mathbb{N}_{n,2}^<$  is an **inversion** of  $\pi$  just in case  $\pi$  “reorders”  $\{i, j\}$  in the sense that  $\pi(i) > \pi(j)$ . This implies that  $(i - j)[\pi(i) - \pi(j)] < 0$  because  $i - j$  and  $\pi(i) - \pi(j)$  have opposite signs.
3. Denote the set of inversions of  $\pi$  by

$$\mathcal{I}(\pi) := \{(i, j) \in \mathbb{N}_{n,2}^< \mid \pi(i) > \pi(j)\}$$

Note that an inversion of  $\pi$  is very different from its inverse!

# The Sign or Parity of a Permutation

## Definition

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

# The Sign of a General Transposition

## Theorem

For each transposition  $\tau_{k,\ell}$  with  $k, \ell \in \mathbb{N}_n$  and  $k < \ell$ :

1. the set of inversions is

$$\mathcal{I}(\tau_{k,\ell}) = \cup_{r \in \mathbb{N} \cap (k,\ell)} \{(k, r), (r, \ell)\} \cup \{(k, \ell)\}$$

where  $\mathbb{N} \cap (k, \ell)$  denotes

the intersection of  $\mathbb{N}$  with the open interval  $(k, \ell)$ ;

2.  $n(\tau_{k,\ell}) = 2(\ell - k) - 1$  and  $\text{sgn}(\tau_{k,\ell}) = -1$ .

## Proof.

If  $\pi$  is the transposition  $\tau_{k,\ell}$ , then  $(k, \ell)$  is evidently an inversion.

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

So if  $i < j$ , then  $\pi(i) \leq i < j \leq \pi(j)$  unless  $i = k$  or  $j = \ell$ .

But then  $(k, r)$  and  $(r, \ell)$  are both inversions iff  $k < r < \ell$ .

The rest of the proof is straightforward. □

# A Two-Part Exercise and a Fundamental Lemma

## Exercise

For each permutation  $\pi \in \Pi_n$ , show that:

1.  $n(\pi) = 0 \iff \pi = \iota$ , the identity permutation;
2.  $n(\pi) = 1$  if and only if  $\pi$  is an adjacency transposition;
3.  $n(\pi) \leq \frac{1}{2}n(n-1)$ , 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

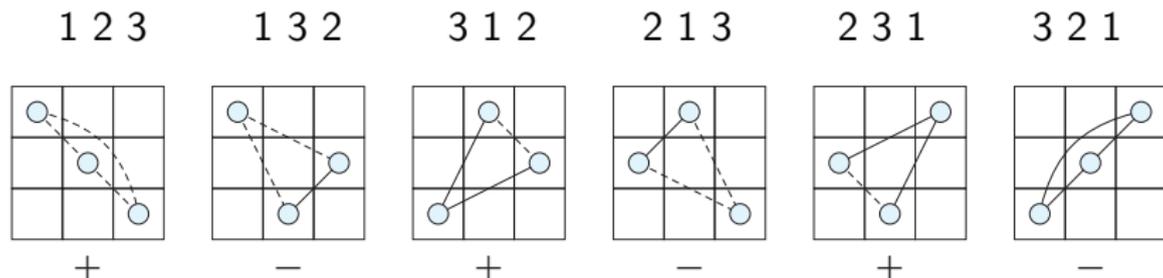
of ordered pairs  $(i, j) \in \mathbb{N}_n \times \mathbb{N}_n$  that satisfy  $i < j$ .

## Lemma

For each adjacency transposition  $\tau_{r,r+1}$  with  $r \in \mathbb{N}_{n-1}$  and each permutation  $\pi \in \Pi_n$ , one has  $\text{sgn}(\pi \circ \tau_{r,r+1}) = -\text{sgn}(\pi)$ .

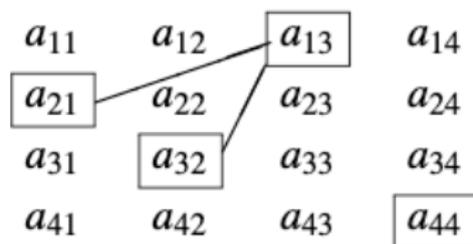
## Examples of the Fundamental Lemma, I

Figure 13.3.1 of EMEA6 and 7, shown below, illustrates the six possible permutations in  $\Pi_3$ , and links between pairs. Each inversion is marked by a solid link; each other link is dashed. Below each permutation is a + or - sign, indicating the parity, which equals the parity of the number of inversions.



Interchanging two adjacent rows replaces any positive dashed link with a negative solid link, and vice versa. This changes the number of inversions by  $\pm 1$ , reversing the parity.

## Examples of the Fundamental Lemma, II



Consider  $\pi \in \Pi_4$  defined so that each boxed element  $a_{ij}$  in the above diagram corresponds to the pair  $(i, \pi(i))$ , for  $i \in \mathbb{N}_4$ .

So the diagram represents the permutation  $\pi^{3124}$ .

It has two inversions  $(1, 2)$  and  $(1, 3)$  where  $j > i$  but  $\pi(j) < \pi(i)$ , implying that  $\text{sgn}(\pi^{3124}) = (-1)^2 = +1$ .

The adjacency transposition  $\tau_{1,2}$  removes the inversion  $(1, 2)$ .

The adjacency transposition  $\tau_{2,3}$  adds the inversion  $(2, 3)$ .

The adjacency transposition  $\tau_{3,4}$  adds the inversion  $(3, 4)$ .

In each case the new matrix  $\pi' = \pi \circ \tau_{r,r+1}$  has sign  $-1$ .

## Examples of the Fundamental Lemma, III

$a_{11}$	$a_{12}$	$a_{13}$	$a_{14}$	$a_{15}$
$a_{21}$	$a_{22}$	$a_{23}$	$a_{24}$	$a_{25}$
$a_{31}$	$a_{32}$	$a_{33}$	$a_{34}$	$a_{35}$
$a_{41}$	$a_{42}$	$a_{43}$	$a_{44}$	$a_{45}$
$a_{51}$	$a_{52}$	$a_{53}$	$a_{54}$	$a_{55}$

Consider  $\pi \in \Pi_5$  defined so that each boxed element  $a_{ij}$  in the above diagram corresponds to the pair  $(i, \pi(i))$ , for  $i \in \mathbb{N}_5$ . So the diagram represents the permutation  $\pi^{23514}$ .

It has four inversions  $(1, 4)$ ,  $(2, 4)$ ,  $(3, 4)$  and  $(3, 5)$  where  $j > i$  but  $\pi(j) < \pi(i)$ , so  $\text{sgn}(\pi^{23514}) = (-1)^4 = +1$ .

The three adjacency transpositions  $\tau_{1,2}$ ,  $\tau_{2,3}$ , and  $\tau_{3,4}$  each remove one inversion — namely  $(1, 2)$ ,  $(2, 3)$ , and  $(3, 4)$ .

The adjacency transposition  $\tau_{4,5}$  adds the inversion  $(4, 5)$ .

In each case the new matrix  $\pi' = \pi \circ \tau_{r,r+1}$  has sign  $-1$ .

## Restatement and Proof of Fundamental Lemma, I

Given permutation  $\pi \in \Pi_n$  and adjacency transposition  $\tau_{r,r+1}$ , we introduce the notation  $\pi' = \pi \circ \tau_{r,r+1}$ .

This implies that  $\pi'(r) = \pi(r+1)$  and  $\pi'(r+1) = \pi(r)$  but  $\pi'(p) = \pi(p)$  and  $\pi'(q) = \pi(q)$  for all  $p < r$  and  $q > r+1$ .

### Lemma

*For each adjacency transposition  $\tau_{r,r+1}$  with  $r \in \mathbb{N}_{n-1}$  and each permutation  $\pi \in \Pi_n$ , one has  $\text{sgn}(\pi') = -\text{sgn}(\pi)$ .*

### Proof.

First, for all  $r \in \mathbb{N}_{n-1}$  and all  $(p, q) \in \mathbb{N}_n \setminus \{r, r+1\}$  with  $p < q$ , one has  $\tau_{r,r+1}(p) = p$  and  $\tau_{r,r+1}(q) = q$ .

This implies that  $(p, q) \in \mathcal{I}(\pi) \iff (p, q) \in \mathcal{I}(\pi')$ .

Second, for the particular pair  $(r, r+1)$  one has

$$\begin{aligned}(r, r+1) \in \mathcal{I}(\pi) &\iff \pi(r) > \pi(r+1) \\ &\iff \pi'(r+1) > \pi'(r) \iff (r, r+1) \notin \mathcal{I}(\pi')\end{aligned}$$

## Proof of Fundamental Lemma, II

### Proof.

To continue the proof, we consider all the four or fewer pairs in  $\{p, q\} \times \{r, r + 1\}$  satisfying  $p < r$  and  $r + 1 < q$ .

1. In case  $p < r$ , one has  $(p, r) \in \mathcal{I}(\pi)$  if and only if

$$\pi(p) = \pi'(p) > \pi(r) = \pi'(r + 1) \iff (p, r + 1) \in \mathcal{I}(\pi')$$

and  $(p, r + 1) \in \mathcal{I}(\pi)$  if and only if

$$\pi(p) = \pi'(p) > \pi(r + 1) = \pi'(r) \iff (p, r) \in \mathcal{I}(\pi')$$

2. In case  $r + 1 < q$ , one has  $(r, q) \in \mathcal{I}(\pi)$  if and only if

$$\pi(r) = \pi'(r + 1) > \pi(q) = \pi'(q) \iff (r + 1, q) \in \mathcal{I}(\pi')$$

and  $(r + 1, q) \in \mathcal{I}(\pi)$  if and only if

$$\pi(r + 1) = \pi'(r) > \pi(q) = \pi'(q) \iff (r, q) \in \mathcal{I}(\pi') \quad \square$$

## Proof of Fundamental Lemma, III

### Proof.

To summarize, given any  $\tau_{r,r+1}$  and any  $\pi \in \Pi_n$ , as well as  $\pi' = \pi \circ \tau_{r,r+1}$ , we have proved that:

1. for each pair  $(p, q) \in \mathbb{N}_n \setminus \{r, r+1\}$ , one has  $(p, q) \in \mathcal{I}(\pi) \iff (p, q) \in \mathcal{I}(\pi')$ ;
2. for each  $p < r$  and  $q > r+1$ , one has:
  - (i)  $(p, r) \in \mathcal{I}(\pi) \iff (p, r+1) \in \mathcal{I}(\pi')$ ;
  - (ii)  $(p, r+1) \in \mathcal{I}(\pi) \iff (p, r) \in \mathcal{I}(\pi')$ ;
  - (iii)  $(r, q) \in \mathcal{I}(\pi) \iff (r+1, q) \in \mathcal{I}(\pi')$ ;
  - (iv)  $(r+1, q) \in \mathcal{I}(\pi) \iff (r, q) \in \mathcal{I}(\pi')$ ;
3.  $(r, r+1) \in \mathcal{I}(\pi) \iff (r, r+1) \notin \mathcal{I}(\pi')$ .

So, apart from the pair  $(r, r+1)$ , any other pair in  $\mathcal{I}(\pi)$  is matched with a unique corresponding pair in  $\mathcal{I}(\pi')$ .

It follows that  $|\mathbf{n}(\pi) - \mathbf{n}(\pi')| = |\#\mathcal{I}(\pi) - \#\mathcal{I}(\pi')| = 1$ , implying that  $\text{sgn}(\pi') \neq \text{sgn}(\pi)$  and so  $\text{sgn}(\pi') = -\text{sgn}(\pi)$ . □

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## Matrices: Introduction

Matrices and Their Transposes

Matrix Multiplication: Definition

## Special Matrices

Square, Symmetric, and Diagonal Matrices

The Identity Matrix

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Transpositions

Adjacency Transpositions

The Inversions and Sign of a Permutation

**The Product Rule**

# The Parity of a Product of Adjacency Transpositions

## Theorem

The sign  $\text{sgn}(\pi^{(k)})$  of the product  $\pi^{(k)} = \prod_{i \in \mathbb{N}_k} \tau_{r_i, r_i+1}$  of any finite collection of  $k$  adjacency transpositions satisfies

$$\text{sgn}(\pi^{(k)}) = \text{sgn} \left( \prod_{i \in \mathbb{N}_k} \tau_{r_i, r_i+1} \right) = (-1)^k$$

## Proof.

We prove the result by induction on  $k$ .

Indeed, when  $k = 1$  one has  $\text{sgn}(\tau_{r_1, r_1+1}) = -1$ .

As the induction hypothesis, suppose  $\text{sgn}(\pi^{(k-1)}) = (-1)^{(k-1)}$  for  $\pi^{(k-1)} = \prod_{i \in \mathbb{N}_{k-1}} \tau_{r_i, r_i+1}$ .

Then the fundamental lemma implies that

$$\text{sgn}(\pi^{(k)}) = \text{sgn} \left( \pi^{(k-1)} \circ \tau_{r_k, r_k+1} \right) = -\text{sgn} \left( \pi^{(k-1)} \right) = (-1)^k \quad \square$$

# The Parity of a General Transposition is Odd

## Theorem

For each pair  $k, \ell \in \mathbb{N}_n$  with  $k < \ell$ ,  
the sign of the transposition  $\tau_{k\ell}$  is  $\text{sgn}(\tau_{k\ell}) = -1$ .

## Proof.

A previous lemma established that the transposition  $\tau_{k\ell}$   
equals both  $\pi^{\ell-1 \searrow k} \circ \pi^{k \nearrow \ell}$  and  $\pi^{k+1 \nearrow \ell} \circ \pi^{\ell \searrow k}$ .

Both of these are the compositions  
of  $2(\ell - k) - 1$  adjacency transpositions.

Because  $2(\ell - k) - 1$  is an odd integer,  
it follows that  $\text{sgn}(\tau_{k\ell}) = (-1)^{2(\ell-k)-1} = -1$ . □

# Products of General Transpositions

## Theorem

For each  $r \in \mathbb{N}$ , the sign  $\text{sgn} \left( \prod_{i \in \mathbb{N}_r} \tau_{k_i, \ell_i} \right)$  of the product of  $r$  general transpositions is  $(-1)^r$ .

## Proof.

For each  $i \in \mathbb{N}_r$ , the transposition  $\tau_{k_i, \ell_i}$  can be expressed as the product of  $2(\ell_i - k_i) - 1$  adjacency transpositions.

So the product  $\prod_{i \in \mathbb{N}_r} \tau_{k_i, \ell_i}$  can be expanded into the product of  $\sum_{i \in \mathbb{N}_r} [2(\ell_i - k_i) - 1]$  adjacency transpositions, whose sign is

$$(-1)^{\sum_{i \in \mathbb{N}_r} [2(\ell_i - k_i) - 1]} = (-1)^{-r} = (-1)^r \quad \square$$

# Products of Permutations

## Theorem

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

## Proof.

Every permutation can be expressed  
as a product of finitely many transpositions.

So suppose that the permutations  $\rho$  and  $\pi$  can be expressed  
as the finite products  $\rho = \prod_{i=1}^k \tau_{r_i, r_{i+1}}$  and  $\pi = \prod_{j=1}^{\ell} \tau_{s_j, s_{j+1}}$   
of  $k$  and  $\ell$  transpositions respectively.

It follows that  $\text{sgn}(\rho) = (-1)^k$  and  $\text{sgn}(\pi) = (-1)^\ell$ .

But  $\rho \circ \pi = \left( \prod_{i=1}^k \tau_{r_i, r_{i+1}} \right) \circ \left( \prod_{j=1}^{\ell} \tau_{s_j, s_{j+1}} \right)$

is the product of  $k + \ell$  transpositions,

so  $\text{sgn}(\rho \circ \pi) = (-1)^{k+\ell} = (-1)^k \cdot (-1)^\ell = \text{sgn}(\rho) \text{sgn}(\pi)$ . □

# The Sign of an Inverse Permutation

## Corollary

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

## Proof.

The identity permutation  $\iota$  has 0 reversals, so  $\text{sgn}(\iota) = (-1)^0 = +1$ .

Then, because  $\iota = \pi \circ \pi^{-1}$ , the product rule implies that

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

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

# A Final Remark

## Example

The same permutation  $\tau_{1,2} \in \Pi_2$  is both:

1. the product  $\tau_{1,2}$  of one transposition;
2. the product  $\tau_{1,2} \circ \tau_{1,2} \circ \tau_{1,2}$  of three transpositions.

## Remark

*The sign of any permutation  $\pi \in \Pi_n$  equals the parity of the number of permutations in any set whose product is  $\pi$ .*

*It is the **parity** of this number that is unique; not the **set** of permutations, nor even the **number** of permutations in this set.*