

# Math 1410, Spring 2020

## Operations on Matrices

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# Overview

1 Recap

2 Transpose

3 Trace

4 Determinants

# Warm-Up

Determine the matrix transformation that:

- 1 Reflects across the  $y$  axis, stretches vertically by a factor of 3, and then rotates by  $45^\circ$ .
- 2 Reflects across the line  $y = x$ , stretches vertically by a factor of 3, then reflects across the  $y$  axis.

# Transpose of a matrix

- The **transpose** of a matrix swaps its rows and columns.
- If  $A = [a_{ij}]$  is an  $m \times n$  matrix, then  $A^T$  is the  $n \times m$  matrix whose  $(i, j)$  entry is  $a_{ji}$ .

Examples: on the board.

# Why transpose?

Why should we care about transpose?

- Admittedly, it's one of several things we teach you to compute in Math 1410, and make you wait until later courses to really understand.
- Later, it gets related to *dual vectors* and *dual operators*, and is important in things like quantum mechanics.
- It gives us an easy way to turn columns into rows and vice versa:

sometimes we write  $[1 \ 2 \ 3]^T$  instead of  $\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$  because it fits better on the page.

- It also connects with the *dot product*: we can write  $\vec{v} \cdot \vec{w} = (\vec{v})^T \vec{w}$ .

# Properties

In each case, assume that  $A$  and  $B$  have the right size for the operation to be defined. (True if both are  $n \times n$  but things work for non-square matrices too.)

- 1  $(A + B)^T = A^T + B^T$
- 2  $(kA)^T = kA^T$
- 3  $(AB)^T = B^T A^T$
- 4  $(A^T)^T = A$
- 5  $(A^T)^{-1} = (A^{-1})^T$ , if  $A$  is invertible.

# Symmetric and antisymmetric matrices

An  $n \times n$  matrix  $A$  is *symmetric* if

$$A^T = A$$

and *antisymmetric* if

$$A^T = -A.$$

What can we say about the entries of  $A$  in each case?

# Constructing symmetric matrices

- 1 Show that  $A + A^T$  is symmetric, and  $A - A^T$  is antisymmetric, for any  $n \times n$  matrix  $A$
- 2 Show that  $AA^T$  and  $A^T A$  are symmetric for any matrix  $A$ .
- 3 Show that any square matrix  $A$  can be written as the sum of a symmetric and an antisymmetric matrix.



## Trace of a matrix

The **trace** of a matrix is simply the sum of its diagonal entries. If  $A = [a_{ij}]$  is an  $n \times n$  matrix, then

$$\operatorname{tr}(A) = \sum_{k=1}^n a_{kk} = a_{11} + a_{22} + \cdots + a_{nn}.$$

(If  $A$  is  $m \times n$  with  $m \neq n$ , sum to whichever of  $m, n$  is smaller.)

Examples? On the board!

## Properties of the trace

- 1  $\text{tr}(A + B) = \text{tr}(A) + \text{tr}(B)$
- 2  $\text{tr}(kA) = k \text{tr}(A)$
- 3  $\text{tr}(AB) = \text{tr}(BA)$  (as long as both products are defined)
- 4  $\text{tr}(A^T) = \text{tr}(A)$

*Note:* on the set of all  $m \times n$  matrices, the pairing

$$\langle A, B \rangle = \text{tr}(B^T A)$$

defines a sort of “dot product”.

# Introduction

The **determinant** is a function that assigns a *number* to any square matrix  $A$ . We denote this number by  $\det A$  or  $|A|$ . We'll define  $\det A$  *recursively*, starting with  $2 \times 2$  matrices, and then showing how to reduce the determinant of a larger matrix to smaller ones. There is a general formula, but it's..... complicated.

## The $2 \times 2$ case

You've already learned how to compute these, back when we did cross products! Given  $A = \begin{bmatrix} a & c \\ b & d \end{bmatrix}$ ,  $\det A = ad - bc$ :

$$\begin{vmatrix} a & c \\ b & d \end{vmatrix} = ad - bc.$$

Note:

- 1 If  $\vec{v} = \begin{bmatrix} a \\ b \end{bmatrix}$  is parallel to  $\vec{v} \begin{bmatrix} c \\ d \end{bmatrix}$ , then  $\det A = 0$ .
- 2 Otherwise,  $\det A$  calculates (up to sign) the area of the parallelogram spanned by  $\vec{v}$  and  $\vec{w}$ .

## The $3 \times 3$ case

This is also not a big leap from cross products:

$$\begin{vmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \end{vmatrix} = a_1 \begin{vmatrix} b_2 & b_3 \\ c_2 & c_3 \end{vmatrix} - a_2 \begin{vmatrix} b_1 & b_3 \\ c_1 & c_3 \end{vmatrix} + a_3 \begin{vmatrix} b_1 & b_2 \\ c_1 & c_2 \end{vmatrix}.$$

You might recall that this is the same as the *scalar triple product*

$\vec{a} \cdot (\vec{b} \times \vec{c})$ . Example: compute the determinant of  $A = \begin{bmatrix} 2 & -1 & 3 \\ 0 & 4 & 1 \\ -1 & 0 & 5 \end{bmatrix}$ .

# Minors and Cofactors

Given an  $n \times n$  matrix  $A = [a_{ij}]$ ,

- The  $(i, j)$  *minor* of  $A$  is the  $(n - 1) \times (n - 1)$  matrix  $M_{ij}$  obtained by deleting row  $i$  and column  $j$  of  $A$
- The  $(i, j)$  *cofactor* of  $A$  is the number  $C_{ij}$  defined by

$$C_{ij} = (-1)^{i+j} \det M_{ij}.$$

Note:  $(-1)^{i+j}$  equals  $+1$  if  $i + j$  is even, and  $-1$  if  $i + j$  is odd. (Now do some examples, Sean)

## Next steps

- In general, we define  $\det A = a_{11}C_{11} + a_{12}C_{12} + \cdots + a_{1n}C_{1n}$  (cofactor expansion along first row). (Each cofactor  $C_{1j}$  is a determinant one smaller than  $A$ )
- Then we prove (OK, state assertively) that we can actually expand any row or column of  $A$ .
- Then we'll observe that determinants of *triangular* matrices are really easy.
- Finally, we'll see what happens to the determinant if we use row operations to get a matrix into triangular form.