$$Sinh(x) = \frac{1}{2}(e^{x} - e^{x})$$

$$\ell^{ix} = \cos(x) + i\sin(x)$$

# Math 381 - Fall 2022 $\tan(z) = \frac{\sin(z)}{\cos(z)}$

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$$f^{(\alpha)}(x, x, y)$$

University of Alberta

Week 7

 $||f^{(\alpha)}(x, y)||_{L^{\infty}(x, y)}$ 
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#### **Last Week**

• We had our midterm exam

#### This Week

• Review of linear algebra

#### **Matrices**

#### **Matrix**

$$A = \begin{bmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,m} \\ a_{2,1} & a_{2,2} & \cdots & a_{2,m} \\ \vdots & \vdots & & \vdots \\ a_{n,1} & a_{n,2} & \cdots & a_{n,m} \end{bmatrix}$$

## **Transpose**

#### Matrix transpose

Given a real matrix  $A \in \mathbb{R}^{n \times m}$ , its transpose  $A^T \in \mathbb{R}^{m \times n}$  has elements

$$A^{T} = \begin{bmatrix} a_{1,1} & a_{2,1} & \cdots & a_{n,1} \\ a_{1,2} & a_{2,2} & \cdots & a_{n,2} \\ \vdots & \vdots & & \vdots \\ a_{1,m} & a_{2,m} & \cdots & a_{n,m} \end{bmatrix}$$

#### Rule

Given two matricies  $A \in \mathbb{R}^{n \times m}$  and  $B \in \mathbb{R}^{m \times q}$ ,

$$(AB)^T = B^T A^T$$
.

Given three matrices

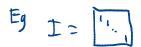
$$(ABC)^T = C^T B^T A^T$$
.

## **Symmetric matrix**

### Symmetric matrix

$$A = A^T$$
 or  $a_{ij} = a_{ji}$ .

## **Diagonal matrices**



#### **Diagonal Matrix**

$$A = \begin{bmatrix} a_{1,1} & 0 & \cdots & 0 \\ 0 & a_{2,2} & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \cdots & a_{n,n} \end{bmatrix}$$

## **Diagonal matrices**

We will omit the zeros when writing matrices in the future.

#### **Diagonal Matrix**

## **Tridiagonal matrices**

#### **Tridiagonal Matrix**

$$A = \begin{bmatrix} a_{1,1} & a_{1,2} & 0 & 0 & \cdots & & 0 \\ a_{2,1} & a_{2,2} & a_{2,3} & 0 & \cdots & & 0 \\ 0 & a_{3,2} & a_{3,3} & a_{3,4} & 0 & \cdots & 0 \\ \vdots & & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & a_{n-1,n-2} & a_{n-1,n-1} & a_{n-1,n} \\ 0 & \cdots & 0 & a_{n,n-1} & a_{n,n} \end{bmatrix}$$

## **Tridiagonal matrices**

We will omit the zeros when writing matrices in the future.

#### **Tridiagonal Matrix**

$$A = \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} & a_{2,3} \\ & a_{3,2} & a_{3,3} & a_{3,4} \\ & & \ddots & \ddots & \ddots \\ & & & & a_{n-1,n} \\ & & & & a_{n,n-1} & a_{n,n} \end{bmatrix}$$

## **Triangular matrices**

#### **Upper Triangular Matrix**

$$A = \begin{bmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n} \\ 0 & a_{2,2} & \cdots & a_{2,n} \\ \vdots & & \ddots & \vdots \\ 0 & 0 & \cdots & a_{n,n} \end{bmatrix}$$

## **Triangular matrices**

We will omit the zeros when writing matrices in the future.

#### **Upper Triangular Matrix**

## Symmetric positive definite matrices

#### Symmetric positive definite matrix

A matrix  $A \in \mathbb{R}^{n \times n}$  is Symmetric positive definite if and only if  $A = A^T$  and

$$x^T A x > 0$$
,  $\forall x \neq 0$ .

## **Orthogonal matrices**

## (Unitary Matrix)

#### Orthogonal matrix

A matrix  $Q \in \mathbb{R}^{n \times n}$  is orthogonal if and only if

$$Q^TQ = I$$



#### 2-norm

$$||x||_2 = \sqrt{\sum_{j=1}^n x_j^2}$$

#### 1-norm

$$||x||_1 = \sum_{j=1}^n |x_j|.$$

#### $\infty$ -norm

$$||x||_{\infty} = \max_{1 \le j \le n} |x_j|.$$

#### p-norm

$$||x||_p = \left(\sum_{j=1}^n x_j^p\right)^{1/p}$$

#### **Definition of a norm**

A norm satisfies three requirements

- **1**  $||x|| \ge 0$  and ||x|| = 0 if and only if x = 0.
- $\|x+y\| \leq \|x\| + \|y\|, \ \forall x,y \in \mathbb{R}^n \quad \text{Triangle inequality}$

#### Matrix norms

#### Induced matrix norm

$$||A|| = \max_{x \neq 0} \frac{||Ax||}{||x||}$$

#### **Submultiplicative property**

$$||AB|| \le ||A|| ||B||$$

## **Spectral Radius**

#### **Spectral radius**

Let  $A \in \mathbb{R}^{n \times n}$  and let  $\Omega$  be the set of eigenvalues of A. The spectral radius is defined as

$$\rho(A) = \max\{|\lambda|; \lambda \in \Omega\}$$

## Eigenvalues and eigenvectors

Given a square matrix  $A \in \mathbb{R}^{n \times n}$ , an eigenvector  $v \in \mathbb{C}^n$  and eigenvalue  $\lambda \in \mathbb{C}$  satisfy

$$Av = \lambda v$$
.

## **Diagonalizable matrices**

A matrix  $A \in \mathbb{R}^{n \times n}$  is diagonalizable if it can be decomposed with

$$A = V \Lambda V^{-1}$$
,

where  $V \in \mathbb{C}^{n \times n}$  is the matrix formed with columns given by the eigenvectors and  $\Lambda \in \mathbb{C}^n$  is a diagonal matrix with diagonal elements given by the eigenvalues.