

# Computational Linear Algebra

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Spring 2022

## 1 Matrices

A table of numbers

### 1.1 Matrix Multiplication Properties

$$(AB)C = A(BC) = ABC$$

$$A(B + C) = AB + AC$$

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

### 1.2 Special Matrices

- Zero Matrix
- Identity Matrix
- Symmetric Matrices  $A^T = A$
- Triagnle matrices or trapezoidal, and diagonal is a subset

#### 1.2.1 Upper triangular matrix

- easy to solve
- eigenvalues are right there
- determinants are right there too

#### 1.2.2 Lower Triangular

if they are both upper triangular and lower triangular, then they are diagonal I don't know much about lower triangular matrices

### 1.2.3 Diagonal Matrices

Very easy to solve a linear system

## 2 Essential Problems of Linear Algebra

### 2.1 Solve Linear Systems

$$Ax = b$$

Use Gauss Jordan Elimination to solve

Computing determinants is  $n!$ , so we essentially never do it

### 2.2 Eigenvalue Problems

Eigenvalues are vectors such that

$$Av = \lambda v$$

bridges are essentially eigenvalue problems

### 2.3 Singular Value Problem

This is about finding vectors  $u$  and  $v$  such that

$$Au = \sigma v$$

Or

$$A^*u = \sigma v$$

### 2.4 Matrix Decomposition

Diagonalization is an example, and SVD is another example. The goal here is to write a matrix as a product of two other matrices. Here are a couple examples

#### 2.4.1 LU

$$A = LU$$

Where  $L$  is lower triangular and  $U$  is upper triangular

Why is this particular composition useful? All the work is done in computing the two matrices  $L$  and  $U$ , and then once you have that, you can solve everything else cheaply. You're essentially done.

#### 2.4.2 QR

This example is

$$A = QR$$

Where  $Q$  is unitary and  $R$  is upper triangular

### 2.4.3 SVD

This one is:

$$A = U\Sigma V^*$$

where  $U$  and  $V^*$  are unitary and  $\Sigma$  is diagonal