



Computational Science:  
Computational Methods in Engineering

# Matrix Terminology & Special Matrices



## Outline

- Matrices & Parts of Matrices
- Special Matrices
- Size & Shape of Matrices
- Health of a Matrix



# Matrices & Parts of Matrices



## Systems of Linear Equations

Systems of equations can be written in matrix form.

$$\begin{array}{l}
 a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1 \\
 a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2 \\
 \vdots \\
 a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n = b_n
 \end{array}
 \rightarrow
 \underbrace{\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & & a_{nn} \end{bmatrix}}_{\substack{[A] \\ \text{or} \\ \mathbf{A}}}
 \underbrace{\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}}_{\substack{[x] \\ \text{or} \\ \mathbf{x}}}
 =
 \underbrace{\begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}}_{\substack{[b] \\ \text{or} \\ \mathbf{b}}}$$



## Rows, Columns, and Diagonals

row 1	$a_{11}$	$a_{12}$	$a_{13}$	$a_{14}$
row 2	$a_{21}$	$a_{22}$	$a_{23}$	$a_{24}$
row 3	$a_{31}$	$a_{32}$	$a_{33}$	$a_{34}$
row 4	$a_{41}$	$a_{42}$	$a_{43}$	$a_{44}$

  

column 1	$a_{11}$	$a_{21}$	$a_{31}$	$a_{41}$
column 2	$a_{12}$	$a_{22}$	$a_{32}$	$a_{42}$
column 3	$a_{13}$	$a_{23}$	$a_{33}$	$a_{43}$
column 4	$a_{14}$	$a_{24}$	$a_{34}$	$a_{44}$

  

diagonal +3	$a_{13}$	$a_{24}$		
diagonal +2	$a_{12}$	$a_{23}$	$a_{34}$	
diagonal +1	$a_{11}$	$a_{22}$	$a_{33}$	$a_{44}$
diagonal 0	$a_{11}$	$a_{22}$	$a_{33}$	$a_{44}$
diagonal -1	$a_{21}$	$a_{32}$	$a_{43}$	
diagonal -2	$a_{31}$	$a_{42}$		
diagonal -3	$a_{41}$			

The center diagonal is usually just called *the diagonal*.

The elements along the center diagonal are sometimes called the *pivot elements*.

## Special Matrices

## Special Matrices (1 of 2)

Symmetric Matrix

$$[A] = \begin{bmatrix} 1 & 2 & 9 & 4 \\ 2 & 6 & 5 & 8 \\ 9 & 5 & 7 & 0 \\ 4 & 8 & 0 & 3 \end{bmatrix}$$

Diagonal Matrix

$$[A] = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 6 & 0 & 0 \\ 0 & 0 & 7 & 0 \\ 0 & 0 & 0 & 3 \end{bmatrix}$$

Identity Matrix

$$[I] = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Zero Matrix

$$[0] = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$



## Special Matrices (2 of 2)

Upper Triangular Matrix

$$[A] = \begin{bmatrix} 1 & 2 & 9 & 4 \\ 0 & 6 & 5 & 8 \\ 0 & 0 & 7 & 5 \\ 0 & 0 & 0 & 3 \end{bmatrix}$$

Triangular matrices can be thought of as "almost" solved matrices. They are very fast to solve.

Lower Triangular Matrix

$$[A] = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 2 & 6 & 0 & 0 \\ 9 & 5 & 7 & 0 \\ 4 & 8 & 1 & 3 \end{bmatrix}$$

Banded Matrix

$$[A] = \begin{bmatrix} 1 & 2 & 0 & 0 \\ 4 & 6 & 5 & 0 \\ 0 & 8 & 7 & 5 \\ 0 & 0 & 10 & 3 \end{bmatrix}$$

Bandwidth of 3

Vandermonde Matrix

$$\begin{bmatrix} 1 & x_1 & x_1^2 & \dots & x_1^N \\ 1 & x_2 & x_2^2 & \dots & x_2^N \\ 1 & x_3 & x_3^2 & \dots & x_3^N \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_{N+1} & x_{N+1}^2 & \dots & x_{N+1}^N \end{bmatrix}$$

Arises when curve fitting to polynomials.  
Usually ill-conditioned for large matrices.



# Block Matrices

Block matrices are “matrices of matrices.”

$$[F] = \begin{bmatrix} [A] & [B] \\ [C] & [D] \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & b_{11} & b_{12} \\ a_{21} & a_{22} & b_{21} & b_{22} \\ c_{11} & c_{12} & d_{11} & d_{12} \\ c_{21} & c_{22} & d_{21} & d_{22} \end{bmatrix}$$

$$[A] = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

$$[B] = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$$

$$[C] = \begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix}$$

$$[D] = \begin{bmatrix} d_{11} & d_{12} \\ d_{21} & d_{22} \end{bmatrix}$$



# Sparse Matrices

Many matrices are composed of almost entirely zeros.

It is not an efficient use of memory to store all these zeros. Instead, only the non-zero elements are stored along with their positions in the matrix.

The opposite of a sparse matrix is a *full matrix*.

$$[A] = \begin{bmatrix} 2 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 2 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 2 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 2 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2 & -1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2 & -1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2 & -1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2 & -1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2 \end{bmatrix}$$



# Size & Shape of Matrices



## Matrix Problem Size

# Equations > # Unknowns

$$\begin{bmatrix} \blacksquare & \blacksquare & \blacksquare \\ \blacksquare & \blacksquare & \blacksquare \\ \blacksquare & \blacksquare & \blacksquare \\ \blacksquare & \blacksquare & \blacksquare \end{bmatrix} \begin{bmatrix} \blacksquare \\ \blacksquare \\ \blacksquare \end{bmatrix} = \begin{bmatrix} \blacksquare \\ \blacksquare \\ \blacksquare \end{bmatrix}$$

Usually occurs when the equations are derived from samples.

Solution is obtained as a *best fit* and is not exact.

Applications

- Curve fitting

# Equations = # Unknowns

$$\begin{bmatrix} \blacksquare & \blacksquare & \blacksquare \\ \blacksquare & \blacksquare & \blacksquare \\ \blacksquare & \blacksquare & \blacksquare \end{bmatrix} \begin{bmatrix} \blacksquare \\ \blacksquare \\ \blacksquare \end{bmatrix} = \begin{bmatrix} \blacksquare \\ \blacksquare \\ \blacksquare \end{bmatrix}$$

Most usual case.

Many standard algorithms exist to obtain an *exact* solution.

Applications

- Circuit theory
- Solving ODEs

# Equations < # Unknowns

$$\begin{bmatrix} \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare \\ \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare \\ \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare \end{bmatrix} \begin{bmatrix} \blacksquare \\ \blacksquare \\ \blacksquare \\ \blacksquare \\ \blacksquare \end{bmatrix} = \begin{bmatrix} \blacksquare \\ \blacksquare \\ \blacksquare \end{bmatrix}$$

Usually occurs when little is known about the problem or solution.

Solution is obtained by *optimization* and is not exact.

Applications

- Topology optimization



# Health of a Matrix



## Health of a Matrix (1 of 3)

Is this system of equations solvable?

$$\begin{array}{l} x+2y+z=8 \\ x+2y+z=8 \\ 3x-y+z=4 \end{array} \quad \begin{bmatrix} 1 & 2 & 1 \\ 1 & 2 & 1 \\ 3 & -1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 8 \\ 8 \\ 4 \end{bmatrix}$$

No!

The 1<sup>st</sup> and 2<sup>nd</sup> equations are the same. The 2<sup>nd</sup> equation does not provide any new information to the problem.

$$\begin{array}{l} x+2y+z=8 \\ 2x+4y+2z=16 \\ 3x-y+z=4 \end{array} \quad \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 3 & -1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 8 \\ 16 \\ 4 \end{bmatrix}$$

No!

The 2<sup>nd</sup> equation is just 2× the 1<sup>st</sup> equation. The 2<sup>nd</sup> equation is still not providing any new information.

$$\begin{array}{l} x+2y+z=8 \\ 4x+y+2z=12 \\ 3x-y+z=4 \end{array} \quad \begin{bmatrix} 1 & 2 & 1 \\ 4 & 1 & 2 \\ 3 & -1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 8 \\ 12 \\ 4 \end{bmatrix}$$

No!

The 2<sup>nd</sup> equation is the sum of the 1<sup>st</sup> and 3<sup>rd</sup> equation, thus the 2<sup>nd</sup> equation still does not provide any new information.



## Health of a Matrix (2 of 3)

Is this system of equations solvable?

$$\begin{array}{l} x+z=8 \\ x+2z=7 \\ 3x+z=4 \end{array} \quad \begin{bmatrix} 1 & 0 & 1 \\ 1 & 0 & 2 \\ 3 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 8 \\ 7 \\ 4 \end{bmatrix}$$

No!

None of these equations contain any information about  $y$ .

So how to know if a problem is solvable?

- All rows must be linearly independent – this ensures they each provide new information to the problem.
- No rows can be all zeros – This would not provide any information.
- No columns can be all zeros – This would be ignoring information from one of the unknowns.

$$\boxed{[A][x] = [b] \text{ is solvable if } \det[A] \neq 0}$$

## Health of a Matrix (3 of 3)

Is the following system of equations solvable?

$$\begin{array}{l} x+2y+z=8 \\ 1.0001x+2y+z=8.0001 \\ 3x-y+z=4 \end{array} \quad \begin{bmatrix} 1 & 2 & 1 \\ 1.0001 & 2 & 1 \\ 3 & -1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 8 \\ 8.0001 \\ 4 \end{bmatrix}$$

Technically yes, but perhaps the solution to be somewhat “touchy” and unstable. This is an *ill-conditioned* matrix.

### Condition Number of a Matrix

The condition number  $\kappa(\mathbf{A})$  of matrix  $\mathbf{A}$  is a measure of how much an answer will change given small changes in the matrix  $\mathbf{A}$ .

Matrices with high condition numbers are less stable. Small changes in the element values of  $\mathbf{A}$  will result in large changes in the elements of  $\mathbf{x}$ .

$$\kappa(\mathbf{A}) = \left| \frac{\sigma_{\max}(\mathbf{A})}{\sigma_{\min}(\mathbf{A})} \right|$$

$\sigma_{\min}(\mathbf{A}) \equiv$  smallest singular value of  $\mathbf{A}$

$\sigma_{\max}(\mathbf{A}) \equiv$  largest singular value of  $\mathbf{A}$



## Example: Condition Number

What is the condition number?

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 1 \\ 1 & 2 & 1 \\ 3 & -1 & 1 \end{bmatrix}$$

$$\kappa(\mathbf{A}) = 5.84 \times 10^{16}$$

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 1 \\ 1 & 0 & 1 \\ 3 & -1 & 1 \end{bmatrix}$$

$$\kappa(\mathbf{A}) = 7.76$$

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 1 \\ 1.0001 & 2 & 1 \\ 3 & -1 & 1 \end{bmatrix}$$

$$\kappa(\mathbf{A}) = 1.4 \times 10^5$$

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 1 \\ 1.01 & 2 & 1 \\ 3 & -1 & 1 \end{bmatrix}$$

$$\kappa(\mathbf{A}) = 1.4 \times 10^3$$