

# Linear Algebra Review

## ORIE 4741

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# Linear Independence and Dependence

# Linear Independence

## Algebraic Definition

### Definition

The sequence of vectors  $v_1, v_2, \dots, v_n$  is **linearly independent** if the only combination that gives the zero vector is  $0v_1 + 0v_2 + \dots + 0v_n$ .

### Example

The vectors  $[1, 0, 0]$ ,  $[0, 1, 0]$ , and  $[0, 0, 1]$  are linearly independent

# Linear Dependence

## Algebraic Definition

### Definition

The sequence of vectors  $v_1, v_2, \dots, v_n$  is **linearly dependent** if there exists a combination that gives the zero vector other than  $0v_1 + 0v_2 + \dots + 0v_n$ .

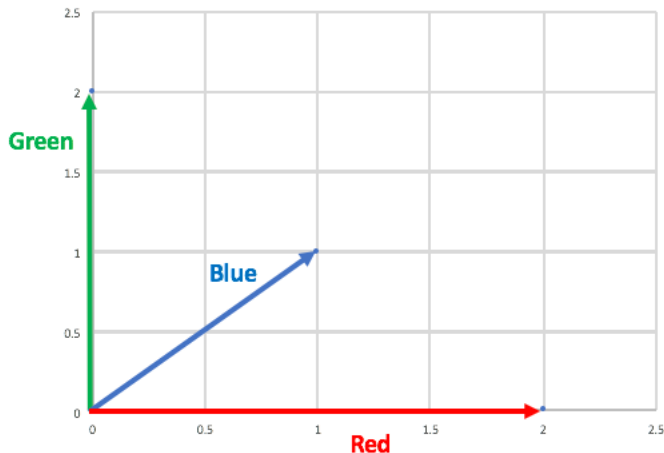
### Example

The vectors  $[1, 2, 0]$ ,  $[2, 4, 0]$ , and  $[0, 0, 1]$  are linearly dependent because  $-2[1, 2, 0] + 1[2, 4, 0] + 0[0, 0, 1] = [0, 0, 0]$ .

# Linear Dependence

## Geometric Definition

Can you find a linear combination of these vectors that equals 0?



Green =  $[0, 2]$

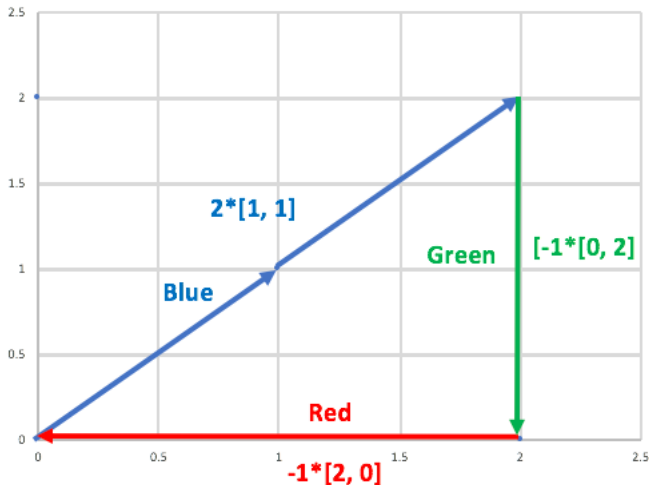
Blue =  $[1, 1]$

Red =  $[2, 0]$

# Linear Dependence

## Geometric Definition

Any multiple of  $-1 * \text{Green} + 2 * \text{Blue} - 2 * \text{Red} = 0$



$$\begin{array}{r} -1 * [0, 2] \\ + 2 * [1, 1] \\ + -1 * [2, 0] \\ \hline [0, 0] \end{array}$$

# Matrix Rank



# Row Rank

## Definition

The **row rank** of a matrix is the number of linearly independent rows in a matrix.

## Example

Each row in matrix A is linearly independent, so  $\text{row rank}(A) = 3$ . Two rows in matrix B are linearly independent, so  $\text{row rank}(B) = 2$ .

$$A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad B = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ -2 & 0 & 0 \end{bmatrix}$$

# Column Rank

## Definition

The **column rank** of a matrix is the number of linearly independent columns in a matrix.

## Example

Each column in matrix A is linearly independent, so  $\text{row rank}(A) = 3$ . The first two columns in matrix B are linearly independent, so  $\text{row rank}(B) = 2$ .

$$A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad B = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ -2 & 0 & 0 \end{bmatrix}$$

Did you notice the row ranks and the column ranks for the matrices were the same?

## Definition

The **rank** of a matrix is the number of linearly independent rows (columns) in a matrix.

# Full-Rank

## Definition

A matrix  $B$  is **full-rank** if:

$$\text{rank}(B) = \min\{\# \text{ columns in } B, \# \text{ of rows in } B\}$$

## Example

Both  $A$  and  $B$  are full-rank.

$$A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \quad B = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

# Invertible Matrices

# Matrix Multiplication

Multiply  $A*B$  where:

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad B = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$

# Matrix Multiplication

Multiply  $A*B$  where:

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad B = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$

$$AB = \begin{bmatrix} (1 * 5) + (2 * 7) & (1 * 6) + (2 * 8) \\ (3 * 5) + (4 * 7) & (3 * 6) + (4 * 8) \end{bmatrix} = \begin{bmatrix} 19 & 22 \\ 43 & 50 \end{bmatrix}$$

# Identity Matrix

## Definition

The **identity matrix**,  $I$ , is an  $n \times n$  matrix with 1s on the diagonal and 0s everywhere else with the property that  $Ix = xI = x$  for any vector  $x$  ( $IA = AI = A$  for any matrix  $A$ ).

## Example

An example of the  $3 \times 3$  identity matrix is:

$$I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$



# Definition of Invertible Matrix

## Definition

The matrix  $A$  is **invertible** if there exists a matrix  $A^{-1}$  such that

- $A^{-1}A = I$
- $AA^{-1} = I$

Note: An invertible matrix should be square (same number of rows and columns) and have full-rank.

# Invertible Matrix Example

$$\text{Let } A = \begin{bmatrix} 4 & 3 \\ 1 & 1 \end{bmatrix}.$$

$$\text{Then, the inverse of } A \text{ is: } A^{-1} = \begin{bmatrix} 1 & -3 \\ 1 & 4 \end{bmatrix}.$$

Check that  $A^{-1}$  satisfies the definition of an inverse:

$$AA^{-1} = \begin{bmatrix} 4 & 3 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & -3 \\ 1 & 4 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$A^{-1}A = \begin{bmatrix} 1 & -3 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} 4 & 3 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ Therefore, } A^{-1} \text{ is the inverse of } A.$$

# Norms

# Vector Norm

## Definition

The **norm** of a vector  $x$ , denoted  $\|x\|$  or  $\|x\|_2$ , is the length of  $x$ :

$$\|x\| = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$$

## Example

$$\text{Let } x = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}, \text{ then}$$

$$\|x\| = \sqrt{(1^2 + 2^2 + 3^2 + 4^2)} = \sqrt{1 + 4 + 9 + 16} = \sqrt{30}$$

# Projection Matrix

## Definition

Let  $A$  be a matrix. The **Aspan** of the columns of  $A$  is the set of all linear combinations of the columns of  $A$ .

## Example

The rows of  $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$  spans all of  $\mathbb{R}^2$ .

# Projection Matrix

## Definition

A square matrix  $P$  is the projection matrix onto the  $\text{span}(\text{columns of } A)$  if

$$P^2 = P.$$

## Example

$P = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ -\frac{1}{2} & \frac{1}{2} \end{bmatrix}$  is a projection matrix because  $P^2 = P$ .

# Projection Matrix Example Continued

## Example

$P = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ -\frac{1}{2} & \frac{1}{2} \end{bmatrix}$  is a projection matrix that projects onto  $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$ .

$$P \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$P \begin{bmatrix} 2 \\ 27 \end{bmatrix} = \begin{bmatrix} -4 \\ 4 \end{bmatrix} = -4 \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$P \begin{bmatrix} 254 \\ -1000 \end{bmatrix} = \begin{bmatrix} 627 \\ -627 \end{bmatrix} = 627 \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$



## Helpful Matrix Hints

# Helpful Matrix Properties and Identities

- $AB \neq BA$
- $A^T B^T = (BA)^T$
- $A(B + C) = AB + AC$
- $(AB)C = A(BC)$
- $IA = AI = A$
- $AA^{-1} = A^{-1}A = I$

# Gradients

# Multivariate Derivatives

Find  $\frac{\partial f}{\partial x}$ ,  $\frac{\partial f}{\partial y}$ ,  $\frac{\partial f}{\partial z}$ , where

$$f(x, y, z) = 5x + 2x^2 + xy + 3xy^2z - 4y^3 + 2yz + 4z^2 - x^4yz^2.$$

# Multivariate Derivatives

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$$\frac{\partial f}{\partial x} = 5 + 4x + y + 3y^2z - 4x^3yz^2$$

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$$\frac{\partial f}{\partial y} = x + 6xyz - 12y^2 + 2z - x^4z^2$$

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$$\frac{\partial f}{\partial x} = 5 + 4x + y + 3y^2z - 4x^3yz^2$$

$$\frac{\partial f}{\partial y} = x + 6xyz - 12y^2 + 2z - x^4z^2$$

$$\frac{\partial f}{\partial z} = 3xy^2 + 2y + 8z - 2x^4yz$$

## Definition

Let  $f(x_1, x_2, \dots, x_n)$  be a multivariable function. The **gradient of  $f$** ,  $\nabla f$ , is the multivariable generalization of the derivative of  $f$ . The gradient is a vector, where each row corresponds to a partial derivative with respect to a variable of the function.

$$\nabla f = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \cdot \\ \cdot \\ \cdot \\ \frac{\partial f}{\partial x_n} \end{bmatrix}$$



# Example Gradient

Let  $f(x, y, z) = x^2 + 3xy + 4xyz + z$ .

Then  $\nabla f$  is:

$$\nabla f = \begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \\ \frac{\partial f}{\partial z} \end{bmatrix} = \begin{bmatrix} 2x + 3y + 4yz \\ 3x + 4xz \\ 4xy + 1 \end{bmatrix}$$

# Computational Complexity Notation

# Big O Notation

Big O notation is used to describe the run-time (computational complexity) of algorithms.

## Definition

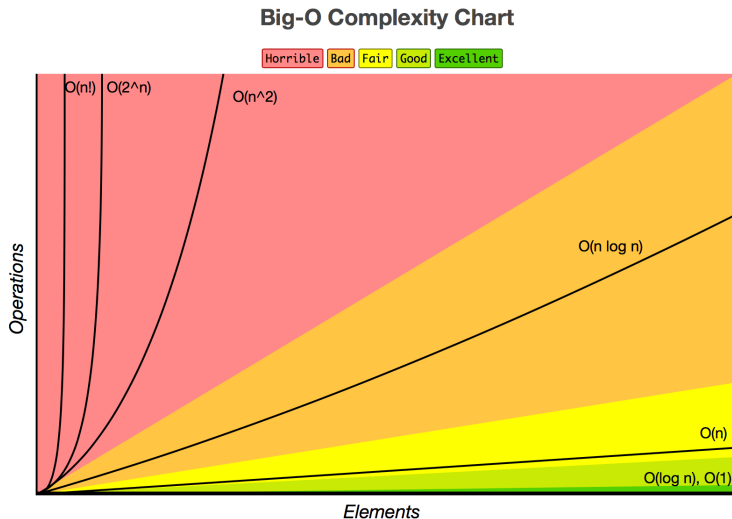
Let  $f(n)$  be the run-time of some algorithm. If  $f(n) = O(g(n))$ , then there exists a constant  $C$  and a constant  $N$  such that:

$$|f(x)| \leq C * |g(x)| \text{ for all } x > N.$$

## Example

When an algorithm runs in time  $O(n)$ . That means it runs in linear time. So, where  $n$  is the amount of data, if an algorithm runs in time  $5n$ ,  $1000n+50000$ ,  $n$ , or  $n-45$ , the algorithm runs in time  $O(n)$ .

# Big O Notation



Strang, Gilbert. (2009, 4th ed.). Introduction to Linear Algebra.

Ling-Hsiao Ly's 2012 Lecture 3 Projection and Projection Matrices Notes

[http://www.ss.ncu.edu.tw/lyu/lecture\\_files/n/lyu\\_LA\\_Notes/Lyu\\_LA\\_2012/Lyu\\_LA\\_32012.pdf](http://www.ss.ncu.edu.tw/lyu/lecture_files/n/lyu_LA_Notes/Lyu_LA_2012/Lyu_LA_32012.pdf)