

Inverse of a Matrix by Determinants

The inverse of a nonsingular $n \times n$ matrix A is given

$$A^{-1} = \frac{1}{\det A} [C_{jk}]^T = \frac{1}{\det A} = \begin{bmatrix} C_{11} & C_{21} & \cdots & C_{n1} \\ C_{12} & C_{22} & & C_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ C_{1n} & C_{2n} & & C_{nn} \end{bmatrix}$$

Example:

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

$$\det A = (1)(2-2) - 2(4-3) + (-1)(4-3) = 0 - 2 - 1 = -3$$

$$\frac{a_{11}}{C_{11}} = 2 - 2 = 0$$

$$\frac{a_{12}}{C_{12}} = -(4-3) = -1$$

$$\frac{a_{13}}{C_{13}} = 4 - 3 = 1$$

$$\frac{a_{21}}{C_{21}} = -(4 - (-2)) = -6$$

$$\frac{a_{22}}{C_{22}} = 2 - (-3) = 5$$

$$\frac{a_{23}}{C_{23}} = -(2-6) = +4$$

$$\frac{a_{31}}{C_{31}} = 2 - (-1) = 3$$

$$\frac{a_{32}}{C_{32}} = -(1 - (-3)) = -4$$

$$\frac{a_{33}}{C_{33}} = 1 - 4 = -3$$

$$A^{-1} = \frac{1}{(-3)} \begin{bmatrix} 0 & -6 & 3 \\ -1 & 5 & 4 \\ 1 & 4 & -3 \end{bmatrix} = \begin{bmatrix} 0 & 2 & -1 \\ 1/3 & -5/3 & 4/3 \\ -1/3 & -4/3 & 1 \end{bmatrix} \quad D = -3$$

$$M = \begin{bmatrix} 0 & 1 & 1 \\ 6 & 5 & -4 \\ 3 & 3 & -3 \end{bmatrix} \rightarrow M^T = \begin{bmatrix} 0 & 6 & 3 \\ 1 & 5 & 3 \\ 1 & -4 & -3 \end{bmatrix} \xrightarrow[\pm]{\frac{M^T}{D}} \begin{bmatrix} 0 & 2 & -1 \\ -1/3 & -5/3 & 4/3 \\ -1/3 & 4/3 & 1 \end{bmatrix}$$

change
sign
 $C_{ij} = -C_{ij}$ if $i+j$ odd
 $= C_{ij}$ even

$$\rightarrow \begin{bmatrix} 0 & 2 & -1 \\ +1/3 & -5/3 & 1 \\ -1/3 & -4/3 & 1 \end{bmatrix} \\ A^{-1}$$

Inverse of a 2×2 Matrix ($\det A \neq 0$)

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

$$\det(\underline{A}) = a_{11}a_{22} - a_{12}a_{21}$$

$$C_{11} = a_{22} \quad C_{12} = -a_{21}$$

$$C_{21} = -a_{12} \quad C_{22} = a_{11}$$

$$\underline{A}^{-1} = \frac{1}{\det(\underline{A})} \begin{bmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{bmatrix}$$

Verification:

$$\begin{aligned} \underline{A} \underline{A}^{-1} &= \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \frac{1}{\det \underline{A}} \begin{bmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{bmatrix} \\ &= \frac{1}{\det \underline{A}} \begin{bmatrix} a_{11}a_{22} - a_{12}a_{21} & -a_{11}a_{12} + a_{11}a_{12} \\ a_{21}a_{22} - a_{21}a_{22} & -a_{12}a_{21} + a_{11}a_{22} \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \end{aligned}$$

Develop a rule for the inverse of a 2×2 matrix with $\det(\underline{A}) \neq 0$

Diagonal Matrices

$\underline{A} = [a_{jk}]$ $a_{jk} = 0$ when $j \neq k$. $\underline{A}$ has an inverse iff all $a_{jj} \neq 0$.

$$\underline{A} = \begin{bmatrix} a_{11} & 0 & \dots & 0 \\ 0 & a_{22} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & a_{nn} \end{bmatrix}$$

Then

$$\underline{A}^{-1} = \begin{bmatrix} 1/a_{11} & 0 & \dots & 0 \\ 0 & 1/a_{22} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1/a_{nn} \end{bmatrix}$$

Properties

$$(\underline{AB})^{-1} = \underline{B}^{-1} \underline{A}^{-1} \quad (\text{In reverse order})$$

Proof

$$(\underline{AB})(\underline{AB})^{-1} = \underline{A} \underbrace{\underline{B} \underline{B}^{-1}}_{\underline{I}} \underline{A}^{-1} = \underline{A} \underline{A}^{-1} = \underline{I}$$

$$\underline{B}(\underline{AB})^{-1} = \underline{B} \underbrace{\underline{B}^{-1}}_{\underline{I}} \underline{A}^{-1} = \underline{A}^{-1}$$

$$(\underline{AB})(\underline{AB})^{-1} = \underline{A} \underline{A}^{-1} = \underline{I}$$

Matrix Multiplication

i. $\underline{A}\underline{B} = \underline{0}$ does not imply $\underline{A} = \underline{0}$ or $\underline{B} = \underline{0}$, or $\underline{B}\underline{A} = \underline{0}$

$$\underline{A} = \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix}, \underline{B} = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$\underline{A}\underline{B} = \begin{bmatrix} -1+1 & 1-1 \\ -2+2 & 2-2 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}, \text{ but } \underline{A} \neq \underline{0} \text{ or } \underline{B} \neq \underline{0}$$

$$\underline{B}\underline{A} = \begin{bmatrix} -1+2 & -1+2 \\ 1-2 & 1-2 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ -1 & -1 \end{bmatrix} \neq \underline{0}$$

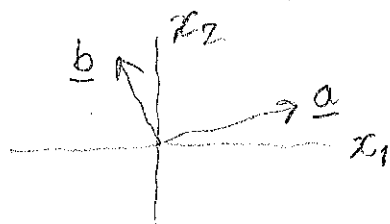
Determinant of a Matrix Product

$$\det(\underline{A}\underline{B}) = \det(\underline{A}) \cdot \det(\underline{B})$$

7.9 Vector spaces

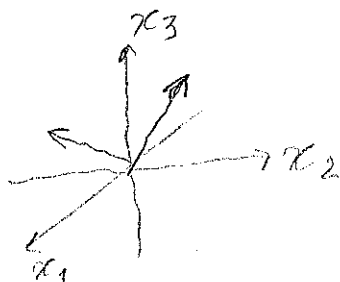
$\mathbb{R}^n$: Real n -dimensional vector space $\mathbb{R}^n$.

$n=2$: Vectors in the plane



$\underline{a}$ & $\underline{b}$ are column or row vectors.
They may be also matrices.

$n=3$: Vectors in 3D-space



Vector addition :

$$\underline{a}, \underline{b} \quad \underline{a} + \underline{b} \quad , \quad \underline{a} \in \mathbb{R}^n, \underline{b} \in \mathbb{R}^n$$

i. Commutativity

$$\underline{a} + \underline{b} = \underline{b} + \underline{a}$$

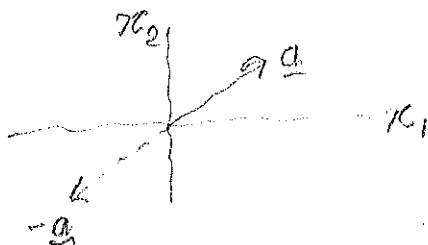
ii. Associativity :

$$(\underline{a} + \underline{b}) + \underline{c} = \underline{a} + (\underline{b} + \underline{c})$$

iii. Zero vector $\underline{0}$ (all components are zero)

$$\underline{a} + \underline{0} = \underline{a}$$

iv. The vector $-\underline{a}$



$$\underline{a} + (-\underline{a}) = \underline{0}$$

Scalar multiplication

c, k : scalar $\underline{a} \in \mathbb{R}^n, \underline{b} \in \mathbb{R}^n$

i. Distributivity:

$$c(\underline{a} + \underline{b}) = c\underline{a} + c\underline{b}$$

$$(c+k)\underline{a} = c\underline{a} + k\underline{a}$$

ii. Associativity

$$c(k\underline{a}) = (ck)\underline{a}$$

$$1\underline{a} = \underline{a}$$

Linear combination:

$$c_1 \underline{a}_1 + c_2 \underline{a}_2 + c_3 \underline{a}_3 + \dots + c_m \underline{a}_m$$

Linearly independent:

$$c_1 \underline{a}_1 + c_2 \underline{a}_2 + \dots + c_m \underline{a}_m = \underline{0}$$

implies that $c_1 = c_2 = \dots = c_m = 0$

If the vector space V has a dimension n or n -dimensional, then it contains a linearly independent set of n vectors.

$n=2$: Two vectors

$n=3$: Three vectors

Vector space of 2×2 matrices.

Consider

$$\underline{B}_{11} = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \underline{B}_{12} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \underline{B}_{21} = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \underline{B}_{22} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

Any 2×2 matrix $\underline{A} = [a_{jk}]$ has a unique representation

$$\underline{A} = a_{11} \underline{B}_{11} + a_{12} \underline{B}_{12} + a_{21} \underline{B}_{21} + a_{22} \underline{B}_{22} = \underline{0} ?$$

iff $a_{11} = a_{12} = a_{21} = a_{22} = 0$. / $\underline{B}_{11}, \underline{B}_{12}, \underline{B}_{21}, \underline{B}_{22}$ basis vectors

$$2D \text{ space} : \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = a \underbrace{\begin{bmatrix} 1 \\ 0 \end{bmatrix}}_{x_1} + b \underbrace{\begin{bmatrix} 0 \\ 1 \end{bmatrix}}_{x_2} \rightarrow \text{column vectors}$$

$\underline{B}_{11}, \underline{B}_{12}, \underline{B}_{21}, \underline{B}_{22}$: Basis vectors

$\underline{x}_1, \underline{x}_2$: Basis vectors

Inner Product Space

Assume $\underline{a}$ and $\underline{b}$ are column vectors.

$$\begin{aligned} \underline{a}^T \underline{b} &= [a_1 \ a_2 \ \dots \ a_n] \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix} \\ &= a_1 b_1 + a_2 b_2 + \dots + a_n b_n \quad (\text{scalar}) \end{aligned}$$

$$\underline{a}^T \underline{b} = (\underline{a}, \underline{b}) = \underline{a} \cdot \underline{b} : \text{Inner product (dot product)}$$

Inner product space

$$\begin{aligned} \text{i. } (q_1 \underline{a} + q_2 \underline{b}, \underline{c}) &= (q_1 \underline{a} + q_2 \underline{b})^T \underline{c} = q_1 \underline{a}^T \underline{c} + q_2 \underline{b}^T \underline{c} \\ &= q_1 (\underline{a}, \underline{c}) + q_2 (\underline{b}, \underline{c}) \end{aligned}$$

$$\text{ii. } (\underline{a}, \underline{b}) = \underline{a}^T \underline{b} = \underline{b}^T \underline{a}$$

$$\text{iii. } (\underline{a}, \underline{a}) \geq 0$$

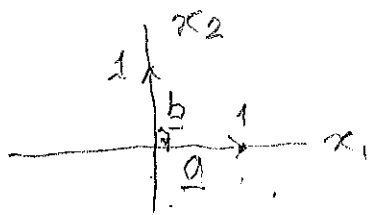
$$\text{iv. } (\underline{a}, \underline{a}) = 0 \text{ iff } \underline{a} = \underline{0}$$

Orthogonal vectors :

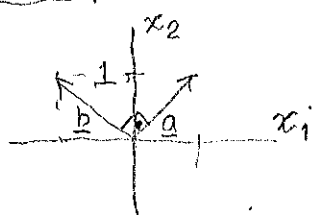
$$(\underline{a}, \underline{b}) = 0 \Rightarrow \underline{a} \text{ \& \& } \underline{b} \text{ are orthogonal}$$

Example : $\underline{a} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \underline{b} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

$$(\underline{a}, \underline{b}) = \underline{a}^T \underline{b} = [1 \ 0] \begin{bmatrix} 0 \\ 1 \end{bmatrix} = 0$$



Example



$$\underline{a} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}; \text{ Determine } \underline{b} \text{ s.t. } \underline{a} \text{ and } \underline{b} \text{ are orthogonal.}$$

$$(\underline{a}, \underline{b}) = 0$$

$$\underline{a}^T \underline{b} = 0 \Rightarrow \begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = 0 \quad b_1 + b_2 = 0 \quad \begin{matrix} b_1 = 1; b_2 = -1 \\ b_1 = -1; b_2 = 1 \end{matrix}$$

Length or norm of a vector

$$\|\underline{a}\| = \sqrt{\underline{a}, \underline{a}} \geq 0$$

If the norm of vector is 1, then $\underline{a}$ is unit vector

Basic Results

i. $|(\underline{a}, \underline{b})| \leq \|\underline{a}\| \|\underline{b}\|$ (Cauchy-Schwarz inequality)

ii. $\|\underline{a} + \underline{b}\| \leq \|\underline{a}\| + \|\underline{b}\|$ (Triangle Inequality)

iii. $\|\underline{a} + \underline{b}\|^2 + \|\underline{a} - \underline{b}\|^2 = 2(\|\underline{a}\|^2 + \|\underline{b}\|^2)$ (Parallelogram equality)

Linear Transformation (Linear Mapping)

$$\underline{y} \text{ \& } \underline{x} \in \mathbb{R}^n$$

$$\left. \begin{array}{l} F(\underline{y} + \underline{x}) = F(\underline{y}) + F(\underline{x}) \\ F(c\underline{x}) = cF(\underline{x}) \end{array} \right\} \text{ Properties of linear transformation}$$

Linear transformation:

$$\underline{x} \in \mathbb{R}^n, \underline{y} \in \mathbb{R}^m; \underline{A} = \text{An } n \times m \text{ matrix}$$

$$\underline{y} = \underline{A} \underline{x} \rightarrow \text{Linear transformation}$$

Example

$$\underline{x} \in \mathbb{R}^n, \underline{y} \in \mathbb{R}^n; \underline{A} = n \times n \text{ nonsingular matrix}$$

$$\underline{y} = \underline{A} \underline{x} \rightarrow \text{Linear transform}$$

$$\underline{A}^{-1} \underline{y} = \underline{A}^{-1} \underline{A} \underline{x} \Rightarrow \underline{x} = \underline{A}^{-1} \underline{y} \text{ (Inverse transform)}$$

8

Matrix Eigenvalue Problems

8.1 The Eigenvalues and Eigenvectors of a Matrix A Assume A is an $n \times n$ nonzero matrix. Consider

$$A \underline{x} = \lambda \underline{x}$$

 λ : Eigenvalues of A $\underline{x}$: Eigenvectors of A Example:

$$A = \begin{bmatrix} 6 & 3 \\ 4 & 7 \end{bmatrix}, \quad \underline{x} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$$

$$A \underline{x} = \begin{bmatrix} 6 & 3 \\ 4 & 7 \end{bmatrix} \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \begin{bmatrix} 18+12 \\ 12+28 \end{bmatrix} = \begin{bmatrix} 30 \\ 40 \end{bmatrix} = 10 \begin{bmatrix} 3 \\ 4 \end{bmatrix}$$

So $\lambda = 10$ is an eigenvalue $\underline{x} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$ is an eigenvectorExample

$$\underline{x} = \begin{bmatrix} 5 \\ 1 \end{bmatrix}$$

$$A \underline{x} = \begin{bmatrix} 6 & 3 \\ 4 & 7 \end{bmatrix} \begin{bmatrix} 5 \\ 1 \end{bmatrix} = \begin{bmatrix} 33 \\ 27 \end{bmatrix} = 3 \begin{bmatrix} 11 \\ 9 \end{bmatrix} \quad \begin{array}{l} \text{No eigenvalue} \\ \text{no eigenvector!} \end{array}$$

Theorem: An $n \times n$ matrix has at least one eigenvalue and at most n numerically different eigenvalues

Definitions $A - \lambda I$: Characteristic matrix $D(\lambda) = \det(A - \lambda I)$: Characteristic polynomial of A $D(\lambda) = 0$: Characteristic equation of A

Solution : $\underline{A} \underline{x} = \lambda \underline{x} = \lambda \underline{I} \underline{x}$ $\underline{I} = n \times n$ identity matrix

$$\underbrace{(\underline{A} - \lambda \underline{I})}_{\text{singular}} \underline{x} = \underline{0}$$

$$\det(\underline{A} - \lambda \underline{I}) = 0$$

$$\text{Let } \underline{A} = \begin{bmatrix} 6 & 3 \\ 4 & 7 \end{bmatrix}, \quad \underline{A} - \lambda \underline{I} = \begin{bmatrix} 6 & 3 \\ 4 & 7 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = \begin{bmatrix} 6-\lambda & 3 \\ 4 & 7-\lambda \end{bmatrix}$$

$$\det(\underline{A} - \lambda \underline{I}) = (6-\lambda)(7-\lambda) - 12$$

$$= 42 - 13\lambda + \lambda^2 - 12 = \lambda^2 - 13\lambda + 30 = 0$$

$$\lambda_{1,2} = \frac{13 \pm \sqrt{169 - 120}}{2} = \frac{13 \pm 7}{2} \begin{matrix} < 10 \\ < 3 \end{matrix}$$

So there are two eigenvalues $\lambda_1 = 10$ & $\lambda_2 = 3$

Eigenvectors :

For $\lambda_1 = 10$:

$$(\underline{A} - 10 \underline{I}) \underline{x} = \underline{0}$$

$$\left(\begin{bmatrix} 6 & 3 \\ 4 & 7 \end{bmatrix} - \begin{bmatrix} 10 & 0 \\ 0 & 10 \end{bmatrix} \right) \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow \begin{bmatrix} -4 & 3 \\ 4 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\left. \begin{array}{l} -4x_1 + 3x_2 = 0 \\ 4x_1 - 3x_2 = 0 \end{array} \right\} \begin{array}{l} 4x_1 = 3x_2 \\ x_2 = 4 \rightarrow x_1 = 3 \end{array} \Rightarrow \underline{x} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$$

check :

$$\underline{A} \underline{x} = \lambda \underline{x} \Rightarrow \begin{bmatrix} 6 & 3 \\ 4 & 7 \end{bmatrix} \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \begin{bmatrix} 18+12 \\ 12+28 \end{bmatrix} = \begin{bmatrix} 30 \\ 40 \end{bmatrix} = \underbrace{10}_{\lambda} \underbrace{\begin{bmatrix} 3 \\ 4 \end{bmatrix}}_{\underline{x}} \quad \checkmark$$

For $\lambda_2 = 3$:

$$(\underline{A} - 3 \underline{I}) \underline{x} = \underline{0} \Rightarrow \left(\begin{bmatrix} 6 & 3 \\ 4 & 7 \end{bmatrix} - \begin{bmatrix} 3 & 0 \\ 0 & 3 \end{bmatrix} \right) \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \underline{0} \Rightarrow \begin{bmatrix} 3 & 3 \\ 4 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\left. \begin{array}{l} 3x_1 + 3x_2 = 0 \\ 4x_1 + 4x_2 = 0 \end{array} \right\} x_1 + x_2 = 0 \Rightarrow x_2 = -x_1 \Rightarrow \underline{x} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

check :

$$\underline{A} \underline{x} = \lambda \underline{x} \Rightarrow \begin{bmatrix} 6 & 3 \\ 4 & 7 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 3 \\ -3 \end{bmatrix} = \underbrace{3}_{\lambda} \underbrace{\begin{bmatrix} 1 \\ -1 \end{bmatrix}}_{\underline{x}} \quad \checkmark$$

Multiple Eigenvalues

Consider the following 3x3 matrix $\underline{A}$

$$\underline{A} = \begin{bmatrix} -2 & 2 & -3 \\ 2 & 1 & -6 \\ -1 & -2 & 0 \end{bmatrix}$$

$$\det(\underline{A} - \lambda \underline{I}) = \begin{vmatrix} -2-\lambda & 2 & -3 \\ 2 & 1-\lambda & -6 \\ -1 & -2 & -\lambda \end{vmatrix}$$

$$= (-2-\lambda) [(1-\lambda)(-\lambda)-12] - 2(-2\lambda-6) - 3(-4+1-\lambda) = 0$$

$$= (-2-\lambda)(-\lambda+\lambda^2-12) - 2(-2\lambda-6) - 3(-3-\lambda) = 0$$

$$= -(2+\lambda)(\lambda^2-\lambda-12) + 4\lambda+12+9+3\lambda = 0$$

$$= -2\lambda^2 + 2\lambda + 24 - \lambda^3 + \lambda^2 + 12\lambda + 7\lambda + 21 = 0$$

$$= -\lambda^3 - \lambda^2 + 21\lambda + 45 = 0$$

$$= -(\lambda^3 + \lambda^2 + 21\lambda + 45) = -(\lambda^3 - 125 + \lambda^2 - 25 - 21(\lambda - 5) - 45 + 125 + 25 - 105)$$

$$= -[(\lambda^3 - 5^3) + (\lambda^2 - 5^2) + 21(\lambda - 5)] = 0$$

$$= -(\lambda - 5)[(\lambda^2 + 5\lambda + 25) + (\lambda + 5) - 21] = 0$$

$$= -(\lambda - 5)[\lambda^2 + 6\lambda + 9] = -(\lambda - 5)(\lambda + 3)^2 = 0$$

$$\lambda_1 = 5, \lambda_2 = \lambda_3 = -3$$

1. Eigenvector for $\lambda_1 = 5$

$$\underline{A} - 5\underline{I} = \begin{bmatrix} -7 & 2 & -3 \\ 2 & -4 & -6 \\ -1 & -2 & -5 \end{bmatrix} \xrightarrow[\substack{7R_2 + R_1 \rightarrow R_2 \\ -7R_3 + R_1 \rightarrow R_3}]{\substack{7R_2 + R_1 \rightarrow R_2 \\ -7R_3 + R_1 \rightarrow R_3}} \begin{bmatrix} -7 & 2 & -3 \\ 0 & -12 & -24 \\ 0 & 16 & 32 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} -7 & 2 & -3 \\ 0 & 1 & 2 \\ 0 & 1 & 2 \end{bmatrix} \rightarrow \begin{bmatrix} -7 & 2 & -3 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{bmatrix} \text{ Rank 2!}$$

$$\left. \begin{aligned} x_2 + 2x_3 &= 0 \Rightarrow x_3 = -1, x_2 = 2 \\ -7x_1 + 2x_2 - 3x_3 &= 0 \Rightarrow -7x_1 + 4 + 3 = 0 \Rightarrow x_1 = 1 \end{aligned} \right\} \underline{x} = \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix}$$

$$\lambda_2 = \lambda_3 = -3$$

$$A - \lambda I = \begin{bmatrix} -2+3 & 2 & -3 \\ 2 & 1+3 & -6 \\ -1 & -2 & -3 \end{bmatrix} = \begin{bmatrix} 1 & 2 & -3 \\ 2 & 4 & -6 \\ -1 & -2 & 3 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} 1 & 2 & -3 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$x_1 + 2x_2 - 3x_3 = 0$$

Generate two vectors:

$$\left. \begin{array}{l} x_2 = 1 \\ x_3 = 0 \end{array} \right\} \begin{array}{l} x_1 + 2 = 0 \Rightarrow x_1 = -2 \\ \Rightarrow \underline{x} = \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix} \end{array} \left. \begin{array}{l} x_2 = 0 \\ x_3 = 1 \end{array} \right\} \begin{array}{l} x_1 + 0 - 3 = 0 \Rightarrow x_1 = 3 \\ \Rightarrow \underline{x} = \begin{bmatrix} 3 \\ 0 \\ 1 \end{bmatrix} \end{array} \right\} \text{Linearly independent!}$$

Algebraic Multiplicity of λ :

The order of M_λ of an eigenvalue λ as a root.

- Above example

For $\lambda = 5 \rightarrow$ algebraic multiplicity is 1.

For $\lambda = -3 \rightarrow$ algebraic multiplicity is 2.

Geometric Multiplicity of λ :

The number of linearly independent eigenvectors to λ .

For $\lambda = 5$, geometric multiplicity is 1

$\lambda = -3$ " " is 2

Example :

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

$$\det(\underline{A} - \lambda \underline{I}) = \begin{vmatrix} -\lambda & 1 \\ 0 & -\lambda \end{vmatrix} = \lambda^2 = 0$$

$$\lambda_1 = \lambda_2 = 0$$

$\Rightarrow$ Algebraic multiplicity is 2. $(M_\lambda = 2)$

Eigenvectors = $\lambda = 0$

$$(\underline{A} - \lambda \underline{I}) \underline{x} = 0 \Rightarrow \underline{A} \underline{x} = 0$$

$$\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \underline{0} \Rightarrow \begin{matrix} x_2 = 0 \\ x_1 = a \end{matrix}$$

Only one eigenvector; $\rightarrow$ geometric multiplicity is 1.
($m_\lambda = 1$)

$$\underline{x} = \begin{bmatrix} a \\ 0 \end{bmatrix}$$

Defect of λ : $\Delta_\lambda = M_\lambda - m_\lambda = 1$!

Transpose :

The transpose $\underline{A}^T$ of a square matrix $\underline{A}$ has the same eigenvalues of $\underline{A}$