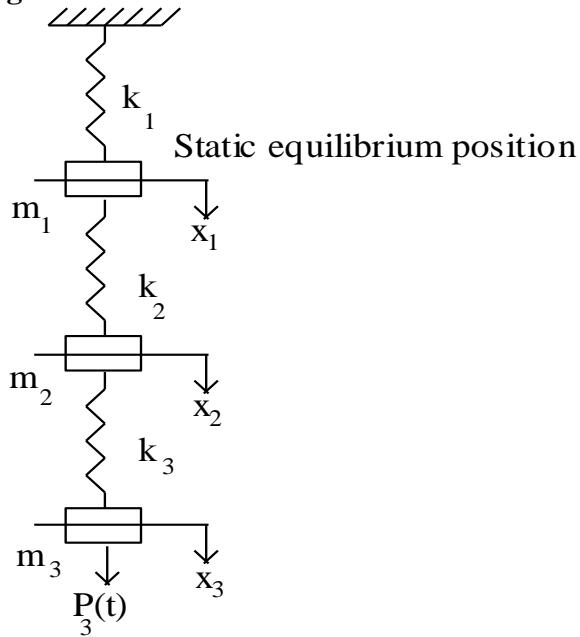


EIGENVALUES & IEGENVECTORS

Eigenvalue Problem:



$$x_1 = x_1(t)$$

$$x_2 = x_2(t)$$

$$x_3 = x_3(t)$$

$$m \tilde{a} = \tilde{F} \text{ for each masses;}$$

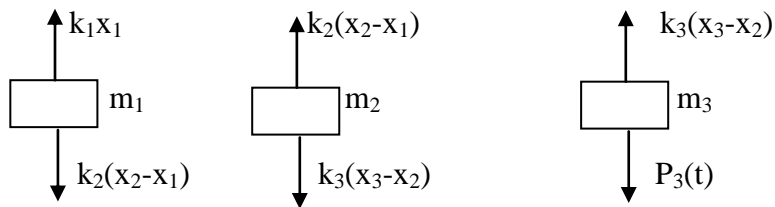
$$a = \ddot{x} = \frac{d^2 x}{dt^2} \quad \begin{aligned} m_1 \ddot{x}_1 &= F_1 \\ m_2 \ddot{x}_2 &= F_2 \\ m_3 \ddot{x}_3 &= F_3 \end{aligned}$$

FBD of masses:

$$\text{Elongation in spring 1} = x_1 \quad F=kx$$

$$\text{Elongation in spring 2} = x_2 - x_1$$

$$\text{Elongation in spring 3} = x_3 - x_2$$



$$m_1 \ddot{x}_1 = k_2(x_2 - x_1) - k_1 x_1 \quad \Rightarrow \quad m_1 \ddot{x}_1 + (k_1 + k_2)x_1 - k_2 x_2 = 0$$

$$m_2 \ddot{x}_2 = k_3(x_3 - x_2) - k_2(x_2 - x_1) \quad \Rightarrow \quad m_2 \ddot{x}_2 - k_2 x_1 + (k_2 + k_3)x_2 - k_3 x_3 = 0$$

$$m_3 \ddot{x}_3 = P_3(t) - k_3(x_3 - x_2) \quad \Rightarrow \quad m_3 \ddot{x}_3 - k_3 x_2 + k_3 x_3 = P_3(t)$$

$$\begin{bmatrix} m_1 & 0 & 0 \\ 0 & m_2 & 0 \\ 0 & 0 & m_3 \end{bmatrix} \begin{bmatrix} \ddot{x}_1 \\ \ddot{x}_2 \\ \ddot{x}_3 \end{bmatrix} + \begin{bmatrix} k_1 + k_2 & -k_2 & 0 \\ -k_2 & k_2 + k_3 & -k_3 \\ 0 & -k_3 & k_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ P_3(t) \end{bmatrix}$$

\underline{M}	$\underline{\ddot{X}}$	\underline{K}	\underline{X}	\underline{P}
Mass	Acc.	Stiffness	Displ.	Applied
Matrix	Matrix	Matrix	Matrix	Force

$$\underline{M} \underline{\ddot{X}} + \underline{K} \underline{X} = \underline{0} \text{ For free vibrations } \underline{P} = \underline{0}$$

Assume the solution:

$$x = a \sin(\omega t + \phi) \text{ where } a : \text{amplitude}$$

ω : angular frequency

ϕ : phase angle

$$\underline{X} = \underline{a} \sin(\omega t + \phi)$$

$$\underline{\dot{X}} = \underline{a} \omega \cos(\omega t + \phi)$$

$$\underline{\ddot{X}} = \underline{a} \omega^2 \sin(\omega t + \phi)$$

$$\underline{M} = [-\omega^2 \underline{a} \sin(\omega t + \phi)] + \underline{K}[\underline{a} \sin(\omega t + \phi)] = \underline{0}$$

$$[-\omega^2 \underline{M} \underline{a} + \underline{K} \underline{a}] \sin(\omega t + \phi) = \underline{0}$$

$$\sin(\omega t + \phi) = 0 \rightarrow \omega t + \phi = 2\pi \rightarrow \phi = 2\pi - \omega t$$

$$\underline{K} \underline{a} = \omega^2 \underline{M} \underline{a} \text{ let } \omega^2 = \lambda$$

$$\boxed{\underline{K} \underline{a} = \lambda \underline{M} \underline{a}} : \text{General Eigenvalue Problem:}$$

λ : Eigenvalue

\underline{a} : Eigenvector

$$\underline{K} \underline{a} = \lambda \underline{M} \underline{a} \quad (*) \text{ GEVP}$$

Premultiply (*) by \underline{M}^{-1}

$$\underline{M}^{-1} \underline{K} \underline{a} = \lambda \underline{a} \text{ let } \underline{A} = \underline{M}^{-1} \underline{K}$$

$$\boxed{\underline{A} \underline{a} = \lambda \underline{a}}$$

Normal eigenvalue problem (NEVP)

Solution of NEVP:

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

$$(\underline{A} - \lambda \underline{I}) \underline{x} = \underline{0} \rightarrow \text{homogeneous equation}$$

In order to have a non trivial solution, $|\underline{A} - \lambda \underline{I}| = 0$

Solution of GEVP:

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

$$(\underline{A} - \lambda \underline{B}) \underline{x} = \underline{0} \rightarrow \text{homogeneous equation for } \underline{x}$$

In order to have a non trivial solution, $|\underline{A} - \lambda \underline{B}| = 0$

Steps to be followed in the solution of an eigenvalue problem.

1. To find the eigenvalues corresponding to $\underline{A}_{n \times n}$

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

where λ (i.e. $\lambda_1, \lambda_2, \dots, \lambda_n$) are eigenvalues

$|\underline{A} - \lambda \underline{I}| = 0 \rightarrow P_n(\lambda)$: characteristic polynomial of \underline{A}

$P_n(\lambda) = 0$ characteristic equation of \underline{A}

$\rightarrow \lambda_1, \lambda_2, \dots, \lambda_n$

2.

$$\underline{A}\underline{x} = \lambda \underline{x} \quad \underline{A}\underline{x}^{(1)} = \lambda_1 \underline{x}^{(1)}$$

$$\underline{A}\underline{x}^{(2)} = \lambda_2 \underline{x}^{(2)}$$

•

•

$$\underline{A}\underline{x}^{(n)} = \lambda_n \underline{x}^{(n)}$$

$$\underline{A}\underline{x}^{(i)} = \lambda_i \underline{x}^{(i)}$$

Example:

$$\underline{A} = \begin{bmatrix} 26 & -2 & 2 \\ 2 & 21 & 4 \\ 4 & 2 & 28 \end{bmatrix} \quad |\underline{A} - \lambda \underline{I}| = 0 \quad \lambda \underline{I} = \begin{bmatrix} \lambda & 0 & 0 \\ 0 & \lambda & 0 \\ 0 & 0 & \lambda \end{bmatrix}$$

$$\begin{vmatrix} 26-\lambda & -2 & 2 \\ 2 & 21-\lambda & 4 \\ 4 & 2 & 28-\lambda \end{vmatrix} = 0 \Rightarrow (\lambda^3 - 75\lambda^2 + 1850\lambda - 15000) = 0 \\ = (\lambda - 20)(\lambda - 25)(\lambda - 30) = 0$$

$$\left. \begin{array}{l} \lambda_1 = 20 \\ \lambda_2 = 25 \\ \lambda_3 = 30 \end{array} \right\} \lambda_1 \langle \lambda_2 \langle \lambda_3$$

$$\underline{A}\underline{x}^{(1)} = 20 \underline{x}^{(1)}$$

$$\underline{A}\underline{x}^{(2)} = 25 \underline{x}^{(2)}$$

$$\underline{A}\underline{x}^{(3)} = 30 \underline{x}^{(3)}$$

$$\begin{bmatrix} 26 & -2 & 2 \\ 2 & 21 & 4 \\ 4 & 2 & 28 \end{bmatrix} \begin{bmatrix} x_1^{(1)} \\ x_2^{(1)} \\ x_3^{(1)} \end{bmatrix} = 20 \begin{bmatrix} x_1^{(1)} \\ x_2^{(1)} \\ x_3^{(1)} \end{bmatrix}$$

$$\begin{bmatrix} 26-20 & -2 & 2 \\ 2 & 21-20 & 4 \\ 4 & 2 & 28-20 \end{bmatrix} \begin{bmatrix} x_1^{(1)} \\ x_2^{(1)} \\ x_3^{(1)} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\left. \begin{aligned} 6x_1 - 2x_2 + 2x_3 &= 0 \\ 2x_1 + x_2 + 4x_3 &= 0 \\ 4x_1 + 2x_2 + 8x_3 &= 0 \end{aligned} \right\} \text{dependent}$$

$$\text{Take an arbitrary } x_3 = \alpha \quad \left. \begin{aligned} 6x_1 - 2x_2 &= -2\alpha \\ 2x_1 + x_2 &= -4\alpha \\ x_2 &= -2\alpha \quad \text{and} \quad x_1 = -\alpha \end{aligned} \right\} \rightarrow \underline{x}^{(1)} = \begin{bmatrix} -\alpha \\ -2\alpha \\ \alpha \end{bmatrix}$$

$$\text{or By GEM: } \begin{bmatrix} 6 & -2 & 2 \\ 2 & 1 & 4 \\ 4 & 2 & 8 \end{bmatrix} \approx \begin{bmatrix} 6 & -2 & 2 \\ 0 & 5/3 & 10/3 \\ 0 & 10/3 & 20/3 \end{bmatrix} \approx \begin{bmatrix} 6 & -2 & 2 \\ 0 & 5/3 & 10/3 \\ 0 & 0 & 0 \end{bmatrix} \rightarrow \underline{x}^{(1)} = \begin{bmatrix} -\alpha \\ -2\alpha \\ \alpha \end{bmatrix}$$

Example GEVP

$$\underline{C} = \begin{bmatrix} 0 & -1 \\ -1 & 2 \end{bmatrix} \quad \underline{B} = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \quad \underline{C} \underline{X} = \lambda \underline{B} \underline{X}$$

$$(\underline{C} - \lambda \underline{B}) \underline{X} = \underline{0} \rightarrow \det(\underline{C} - \lambda \underline{B}) = 0$$

$$\lambda \underline{B} = \begin{bmatrix} 2\lambda & \lambda \\ \lambda & 2\lambda \end{bmatrix} \quad \underline{C} - \lambda \underline{B} = \begin{bmatrix} (-2\lambda) & (-1-\lambda) \\ (-1-\lambda) & (2-2\lambda) \end{bmatrix}$$

$$\begin{aligned} \det(\underline{C} - \lambda \underline{B}) &= -2\lambda(2-2\lambda) - (-1-\lambda)(-1-\lambda) \\ &= -4\lambda + 4\lambda^2 - 1 - 2\lambda - \lambda^2 = 3\lambda^2 - 6\lambda - 1 \end{aligned}$$

$$\lambda_1 = -0.1547$$

$$\lambda_2 = 2.1547$$

$$\underline{C} \underline{X}^{(1)} = \lambda_1 \underline{B} \underline{X}^{(1)}$$

$$\begin{bmatrix} 0 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} x_1^{(1)} \\ x_2^{(1)} \end{bmatrix} = -0.15 \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1^{(1)} \\ x_2^{(1)} \end{bmatrix}$$

$$\left. \begin{aligned} -x_2^{(1)} &= -0.30x_1^{(1)} - 0.15x_2^{(1)} && :0.30x_1 - 0.85x_2 = 0 \\ x_1^{(1)} + 2x_2^{(1)} &= -0.15x_1^{(1)} - 0.30x_2^{(1)} && : -0.85x_1 + 2.30x_2 = 0 \end{aligned} \right\} \text{dependent}$$

let $x_1 = \alpha$

$$x_2 = -\frac{0.85}{0.230}\alpha = -0.37\alpha \quad \underline{x}^{(1)} = \begin{bmatrix} \alpha \\ -0.37\alpha \end{bmatrix}$$

$$\alpha^2 + 0.137\alpha^2 = 1 \quad \alpha = \pm 0.94$$

$$\underline{x}^{(1)} = \pm 0.94 \begin{bmatrix} 1 \\ -0.37 \end{bmatrix} \rightarrow \text{normalized eigenvector. (Normalization wrt length of a vector.)}$$

$$\underline{C} \underline{X}^{(2)} = \lambda_2 \underline{B} \underline{X}^{(2)}$$

$$\begin{bmatrix} 0 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} x_1^{(2)} \\ x_2^{(2)} \end{bmatrix} = 2.15 \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1^{(2)} \\ x_2^{(2)} \end{bmatrix}$$

$$-x_2^{(2)} = 4.30x_1^{(2)} + 2.15x_2^{(2)}$$

$$-x_1^{(2)} + 2x_2^{(2)} = 2.15x_1^{(2)} + 4.30x_2^{(2)}$$

$$\left. \begin{aligned} 4.30x_1 + 3.15x_2 &= 0 \\ 3.15x_1 + 2.30x_2 &= 0 \end{aligned} \right\} \text{dependent}$$

let $x_1 = \alpha$

$$x_2 = -\frac{3.15}{4.30}\alpha = -0.73\alpha \quad \underline{x}^{(2)} = \begin{bmatrix} \alpha \\ -0.73\alpha \end{bmatrix}$$

$$\alpha^2 + 0.53\alpha^2 = 1 \quad \alpha = \pm 0.81$$

$$\underline{x}^{(2)} = \pm 0.81 \begin{bmatrix} 1 \\ -0.73 \end{bmatrix} \rightarrow \text{normalized eigenvector. (Normalization wrt length of a vector.)}$$

Properties of eigenvalues and eigenvectors:

Def 1: A matrix \underline{A} is said to be positive definite if $\underline{x}^T \underline{A} \underline{x} > 0$ for an arbitrary vector \underline{x} .

Remark : if components of a vector can be written as: $x_1^2 + x_2^2 + x_3^2 = 1 \Rightarrow$ positive definite in quadratic forms. Unit vectors are always positive definite quadratic forms.

Def 2: Eigenvectors are said to be orthogonal if they satisfy $\underline{x}^{(i)T} \underline{x}^{(j)} = 0$ for $i \neq j$.

Def 3: Eigenvectors are said to be orthonormal (mutually perpendicular and unit vectors); if they satisfy: $\underline{x}^{(i)T} \underline{x}^{(j)} = \delta_{ij}$

Def 4: Eigenvectors are said to be orthogonal wrt. a matrix \underline{C} , if they satisfy:

$$\underline{x}^{(i)T} \underline{C} \underline{x}^{(j)} = 0 \quad i \neq j$$

Theorem Given a square matrix \underline{A} which is:

- Real
- Symmetric
- Positive definite

Then:

- i) The eigenvalues of \underline{A} are real
- ii) The eigenvectors are orthogonal
- iii) The eigenvalues of \underline{A} are positive

Diagonal form of a matrix:

Def : Modal Matrix $\underline{Q} = [\underline{x}^{(1)} \quad \underline{x}^{(2)} \quad \dots \quad \underline{x}^{(n)}]$ $\underline{x}^{(i)}$: are eigenvectors

If \underline{D} is the diagonal form of \underline{A} , then,

$$\underline{D} = \underline{Q}^{-1} \underline{A} \underline{Q} = \begin{bmatrix} \lambda_1 & 0 & 0 & 0 \\ 0 & \lambda_2 & 0 & 0 \\ \cdot & \cdot & \cdot & \cdot \\ 0 & 0 & 0 & \lambda_n \end{bmatrix}$$

Example :

$$\underline{A} = \begin{bmatrix} 10 & -6 & 0 \\ -6 & 10 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \begin{matrix} \lambda_1 = 1 \\ \lambda_2 = 4 \\ \lambda_3 = 16 \end{matrix} \quad \underline{x}^{(1)} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \quad \underline{x}^{(2)} = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \\ 0 \end{bmatrix} \quad \underline{x}^{(3)} = \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \\ 0 \end{bmatrix}$$

$$\underline{Q} = \begin{bmatrix} 0 & 1/\sqrt{2} & 1/\sqrt{2} \\ 0 & 1/\sqrt{2} & -1/\sqrt{2} \\ 1 & 0 & 0 \end{bmatrix} \quad \underline{Q}^{-1} = \begin{bmatrix} 0 & 0 & 1 \\ 1/\sqrt{2} & 1/\sqrt{2} & 0 \\ 1/\sqrt{2} & -1/\sqrt{2} & 0 \end{bmatrix}$$

$$\underline{D} = \begin{bmatrix} 0 & 0 & 1 \\ 1/\sqrt{2} & 1/\sqrt{2} & 0 \\ 1/\sqrt{2} & -1/\sqrt{2} & 0 \end{bmatrix} \begin{bmatrix} 10 & -6 & 0 \\ -6 & 10 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1/\sqrt{2} & 1/\sqrt{2} \\ 0 & 1/\sqrt{2} & -1/\sqrt{2} \\ 1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 16 \end{bmatrix}$$

Theorem : If \underline{A} is a symmetric matrix, then the modal matrix \underline{Q} is orthogonal.

$$\underline{Q}^{-1} = \underline{Q}^T$$

Square root of a matrix:

let \underline{D} be a diagonal matrix:

$$\underline{D} = \begin{bmatrix} d_{11} & 0 & 0 & \dots & 0 \\ 0 & d_{22} & & & \\ 0 & & d_{33} & & \\ \cdot & & & \cdot & \\ \cdot & & & \cdot & \\ 0 & & & & d_{nn} \end{bmatrix}$$

$$\underline{D}^2 = \underline{D}\underline{D} = \begin{bmatrix} d_{11} & 0 & 0 & \dots & 0 \\ 0 & d_{22} & & & \\ 0 & & d_{33} & & \\ \cdot & & & \cdot & \\ \cdot & & & \cdot & \\ 0 & & & & d_{nn} \end{bmatrix} \begin{bmatrix} d_{11} & 0 & 0 & \dots & 0 \\ 0 & d_{22} & & & \\ 0 & & d_{33} & & \\ \cdot & & & \cdot & \\ \cdot & & & \cdot & \\ 0 & & & & d_{nn} \end{bmatrix} = \begin{bmatrix} d_{11}^2 & 0 & 0 & \dots & 0 \\ 0 & d_{22}^2 & & & \\ 0 & & d_{33}^2 & & \\ \cdot & & & \cdot & \\ \cdot & & & \cdot & \\ 0 & & & & d_{nn}^2 \end{bmatrix}$$

$$\text{then, } \sqrt{D} = \begin{bmatrix} \sqrt{d_{11}} & 0 & 0 & \dots & 0 \\ 0 & \sqrt{d_{22}} & & & \\ 0 & & \sqrt{d_{33}} & & \\ \vdots & & & \ddots & \\ \vdots & & & & \sqrt{d_{nn}} \\ 0 & & & & \sqrt{d_{nn}} \end{bmatrix}$$

$$\text{for any matrix } \underline{A}, \quad \underline{A}^{1/2} = \underline{Q} \underline{D}^{1/2} \underline{Q}^{-1} \quad \underline{A}^{1/2} \underline{A}^{1/2} = \underline{A}$$

$$\text{In general } \underline{A}^{1/n} = \underline{Q} \underline{D}^{1/n} \underline{Q}^{-1} \quad \text{or} \quad \underline{A}^n = \underline{Q} \underline{D}^n \underline{Q}^{-1}$$

Example:

$$\underline{A} = \begin{bmatrix} 10 & -6 & 0 \\ -6 & 10 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \underline{Q} = \begin{bmatrix} 0 & 1/\sqrt{2} & 1/\sqrt{2} \\ 0 & 1/\sqrt{2} & -1/\sqrt{2} \\ 1 & 0 & 0 \end{bmatrix}$$

$$\underline{Q}^{-1} = \underline{Q}^T \quad \underline{D} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 16 \end{bmatrix}$$

$$\underline{A}^{1/2} = \underline{Q} \underline{D}^{1/2} \underline{Q}^{-1}$$

$$\underline{A}^{1/2} = \begin{bmatrix} 0 & 1/\sqrt{2} & 1/\sqrt{2} \\ 0 & 1/\sqrt{2} & -1/\sqrt{2} \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 4 \end{bmatrix} \begin{bmatrix} 0 & 0 & 1 \\ 1/\sqrt{2} & 1/\sqrt{2} & 0 \\ 1/\sqrt{2} & -1/\sqrt{2} & 0 \end{bmatrix}$$

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

Note : trace of $\underline{A} \equiv \text{tr}(\underline{A}) = \sum_{i=1}^n a_{ij} = a_{11} + a_{22} + \dots + a_{nn}$