

Eigenproblems

2x2 Case:
Linear Oscillator

simple oscillator

$$m\ddot{u} + c\dot{u} + ku = 0$$

$$u(0) = u_0, \dot{u}(0) = \dot{u}_0$$

$$\Downarrow \quad w_1 = u, w_2 = \dot{u}$$

$$\begin{pmatrix} \frac{dw_1}{dt} \\ \frac{dw_2}{dt} \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ -\frac{k}{m} & -\frac{c}{m} \end{pmatrix} \begin{pmatrix} w_1 \\ w_2 \end{pmatrix} \quad \text{or} \quad \underbrace{\begin{pmatrix} \frac{dw_1}{dt} \\ \frac{dw_2}{dt} \end{pmatrix}}_{\frac{dw}{dt}} = \underbrace{\begin{pmatrix} 0 & 1 \\ -\omega_n^2 & -2\zeta\omega_n \end{pmatrix}}_A \underbrace{\begin{pmatrix} w_1 \\ w_2 \end{pmatrix}}_w$$

$$\omega_n^2 = \frac{k}{m}, \quad \zeta = \frac{c}{2m\omega_n} \quad (\zeta < 1: \text{underdamped})$$

Assume solution(s) of form

$$w = \underbrace{X}_{2 \times 1} \cdot \underbrace{e^{\lambda t}}_{1 \times 1} \quad \begin{array}{l} X: \text{eigenvector (or mode)} \\ \lambda: \text{eigenvalue} \end{array}$$

$$= \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} e^{\lambda t} = \begin{pmatrix} x_1 e^{\lambda t} \\ x_2 e^{\lambda t} \end{pmatrix} \begin{array}{l} \leftarrow w_1(t) \\ \leftarrow w_2(t) \end{array}$$

Then

$$\frac{dw}{dt} = \begin{pmatrix} \frac{dw_1}{dt} \\ \frac{dw_2}{dt} \end{pmatrix} = \begin{pmatrix} x_1 \lambda e^{\lambda t} \\ x_2 \lambda e^{\lambda t} \end{pmatrix} = \lambda \begin{pmatrix} x_1 e^{\lambda t} \\ x_2 e^{\lambda t} \end{pmatrix} = \lambda X e^{\lambda t}$$

and

$$Aw = Ax e^{\lambda t}$$

Hence

$$\frac{dw}{dt} = Aw$$

$$\begin{aligned} \text{I} \rightarrow x \lambda e^{\lambda t} &= Ax e^{\lambda t}, \text{ or} \\ (Ax - \lambda Ix) e^{\lambda t} &= 0, \text{ or} \\ (A - \lambda I)x &= 0 \end{aligned}$$

$$\Rightarrow x = 0 \text{ or } A - \lambda I \text{ is singular } (\Rightarrow x \neq 0)$$

λ such that

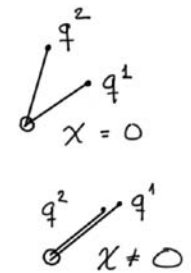
For our oscillator,

$x \neq 0$

$$\left(\begin{pmatrix} 0 & 1 \\ -\omega_n^2 & -2s\omega_n \end{pmatrix} - \begin{pmatrix} \lambda & 0 \\ 0 & \lambda \end{pmatrix} \right) \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0$$

$$\Rightarrow \begin{pmatrix} -\lambda & 1 \\ -\omega_n^2 & -2s\omega_n - \lambda \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$\Rightarrow x_1 \underbrace{\begin{pmatrix} -\lambda \\ -\omega_n^2 \end{pmatrix}}_{q^1} + x_2 \underbrace{\begin{pmatrix} 1 \\ -2s\omega_n - \lambda \end{pmatrix}}_{q^2} = 0$$



Hence to obtain non-trivial x ,

$$\det(\lambda - \lambda I) = p(\lambda)$$

easy way

$$q^2 = \alpha q^1, \text{ or}$$

$$\begin{pmatrix} 1 \\ -2s\omega_n - \lambda \end{pmatrix} = \alpha \begin{pmatrix} -\lambda \\ -\omega_n^2 \end{pmatrix} = \begin{pmatrix} -\alpha\lambda \\ -\alpha\omega_n^2 \end{pmatrix}, \text{ or}$$

$$\begin{aligned} 1 = -\alpha\lambda &\Rightarrow \alpha = -1/\lambda \quad \checkmark \quad x = \begin{pmatrix} -\alpha \\ 1 \end{pmatrix} \cdot \text{const} \\ -2s\omega_n - \lambda = -\alpha\omega_n^2 &\leftarrow 1/\lambda \omega_n^2 \quad \text{"one-handed"} \end{aligned}$$

$$\Rightarrow \lambda^2 + 2s\omega_n\lambda + \omega_n^2 = 0 \quad p(\lambda) = 0$$

$p(\lambda) = \lambda^2 + 2s\omega_n\lambda + \omega_n^2$ is characteristic polynomial

\Rightarrow roots $\lambda = \lambda_1, \lambda = \lambda_2$ eigenvalues

$\Rightarrow (A - \lambda_1)x^1 = 0, (A - \lambda_2)x^2 = 0$ eigenvector. *pairs*

For our oscillator,

$s < 1$

$$p(\lambda) = \lambda^2 + 2s\omega_n\lambda + \omega_n^2 = 0$$

$$\lambda = \frac{-2s\omega_n \pm \sqrt{4s^2\omega_n^2 - 4\omega_n^2}}{2}$$

$$= \frac{-2s\omega_n \pm 2i\omega_n\sqrt{1-s^2}}{2}$$

$$\lambda_1 = -s\omega_n + i\omega_n\sqrt{1-s^2} \Rightarrow x^1$$

$$\lambda_2 = -s\omega_n - i\omega_n\sqrt{1-s^2} \Rightarrow x^2$$

Thus

initial conditions $\Rightarrow c_1, c_2$

$$u(t) = c_1 X^1 e^{\lambda_1 t} + c_2 X^2 e^{\lambda_2 t},$$

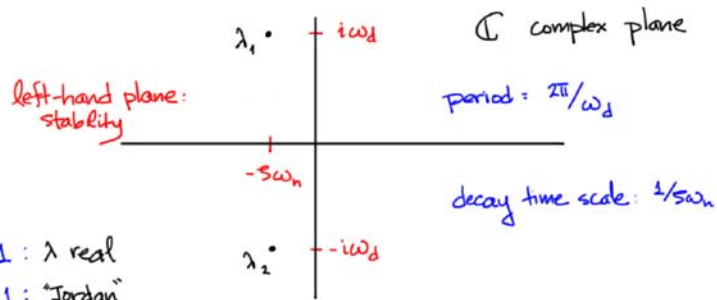
which is a decaying sinusoid ($\zeta < 1$: underdamped):

$$\lambda_{1,2} = -\zeta \omega_n \pm i \omega_n \sqrt{1 - \zeta^2} \quad ; \quad e^{\lambda_{1,2} t} = e^{\text{Re } \lambda t} e^{i \text{Im } \lambda t}$$

$n \times n$ Case:
String in Tension

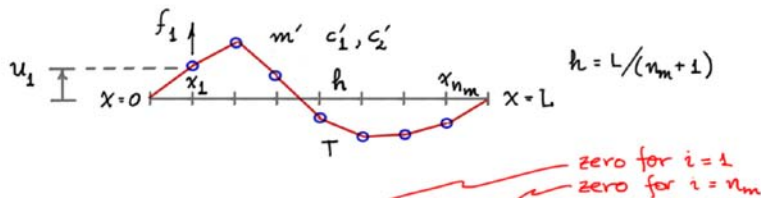
lambdas = eig(A) eig(λ , eye(n,n))
lambdas = eig(\bar{A} , M)

(eigs)



$\zeta > 1$: λ real
 $\zeta = 1$: "Jordan"

Initial Value Problem



$$\begin{cases} m' \ddot{u}_i + c'_1 \dot{u}_i + c'_2 T \left(\frac{-\dot{u}_{i-1} + 2\dot{u}_i - \dot{u}_{i+1}}{h^2} \right) \\ + T \left(\frac{-u_{i-1} + 2u_i - u_{i+1}}{h^2} \right) = f_i, \\ u_i(0) = (u_0)_i, \quad \dot{u}_i(0) = (\dot{u}_0)_i \end{cases} \quad 1 \leq i \leq n_m.$$

zero for $i=1$
zero for $i=n_m$

State variables:

$$w_1 = u_1, w_2 = \dot{u}_1, w_3 = u_2, w_4 = \dot{u}_2, \dots, w_n \text{ for } n = 2n_m.$$

State equations:

$$\begin{aligned} m' \dot{w}_1 &= m' w_2 \\ m' \dot{w}_2 &= -c'_1 w_2 - c'_2 \frac{T}{h^2} (2w_2 - w_4) - \frac{T}{h^2} (2w_1 - w_3) + f' \\ m' \dot{w}_3 &= m' w_4 \\ m' \dot{w}_4 &= -c'_2 w_4 - \dots \\ &\vdots \end{aligned}$$

$$\Rightarrow \begin{cases} M \frac{dw}{dt} = \bar{A} w + \bar{F}, \quad 0 < t \leq t_f \\ w(0) = w_0 \end{cases}$$

Eigenproblem

Insert $w(t) = X e^{\lambda t}$ into

$$M \dot{w} = \bar{A} w \text{ to obtain}$$

$$M \lambda X e^{\lambda t} = \bar{A} X e^{\lambda t}$$

or

$$\boxed{\bar{A} X = \lambda M X}$$

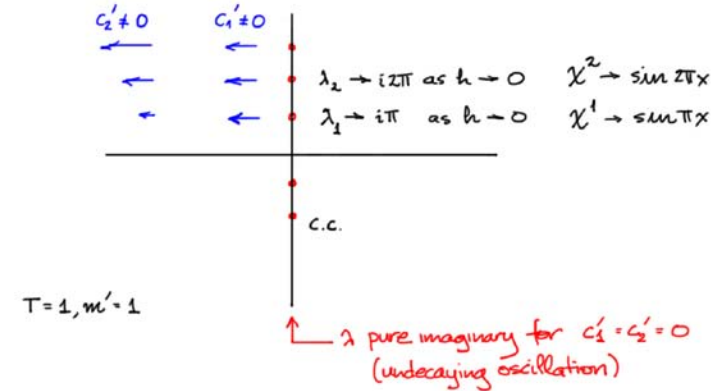
"generalized"

$\det(A - \lambda M) = 0$, OR
columns of $A - \lambda M$ linearly dependent

\downarrow
 (λ_i, X^i) pairs, $1 \leq i \leq n$.

Note: X^i may not be linearly independent, but "typically"...

Spectrum:



$$e^{\lambda t} = e^{\operatorname{Re}(\lambda)t} e^{i \operatorname{Im}(\lambda)t} \quad \left\{ \begin{array}{l} \text{decay time scale } \frac{1}{|\operatorname{Re}(\lambda)|} \\ \text{period } \frac{2\pi}{|\operatorname{Im}(\lambda)|} \end{array} \right.$$

Applications of Eigenproblems:

Linear Stability Analysis \leftarrow Pendulum

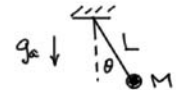
Resonance (Avoidance)

Modal Analysis

⋮

(to dynamics...)

0) formulate problem



$$\begin{cases} \frac{dw_1}{dt} = w_2 & w_1(0) = \theta_0 \\ \frac{dw_2}{dt} = -d_1 w_2 - d_2 |w_2| w_2 - \frac{g_0}{L} \sin(w_1) & w_2(0) = \dot{\theta}_0 \end{cases}$$

1) find equilibria: $\frac{d\bar{w}}{dt} = 0$

$$\underbrace{\begin{pmatrix} \bar{w}_1 \\ \bar{w}_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \bar{w}_1 \\ \bar{w}_2 \end{pmatrix} = \begin{pmatrix} \pi \\ 0 \end{pmatrix}, \dots}_{\text{analyze here}}$$

2) linearize about (each) equilibrium:

Insert

$$w_1 = 0 + w_1' \quad w_2 = 0 + w_2'$$

into state equations to obtain

$$\frac{d(0 + w_1')}{dt} = (0 + w_2') \quad , \quad w_1(0) = \theta_0'$$

$$\frac{d(0 + w_2')}{dt} = -d_1(0 + w_2') - d_2 |0 + w_2'| (0 + w_2') - \frac{g_2}{L} \sin(0 + w_1') \quad , \quad w_2(0) = \dot{\theta}_0'$$

assume $|w_1'| \ll 1, |w_2'| \ll 1$

$$\Rightarrow \left. \begin{aligned} |w_2'| w_2' &\ll |w_2'| \\ \sin(w_1') &= w_1' + O(w_1'^3) \end{aligned} \right\}$$

$$\Rightarrow \begin{cases} \frac{dw_1'}{dt} = w_2' \\ \frac{dw_2'}{dt} = \frac{-g_2}{L} w_1' - d_1 w_2' \end{cases} \quad , \quad \text{or}$$

$$\text{Id} \frac{dw'}{dt} = \underbrace{\begin{pmatrix} 0 & -1 \\ \frac{-g_2}{L} & -d_1 \end{pmatrix}}_A \begin{pmatrix} w_1' \\ w_2' \end{pmatrix}$$

3) pose eigenproblem: $w' = \lambda w$

$$A w = \lambda w$$

$$\lambda \text{Id} w$$

4) find eigenvalues:

$$\text{lambdas} = \text{eig}(A)$$

$$[\text{generalized: lambdas} = \text{eig}(\bar{A}, M)]$$

here $\det(A - \lambda I) \Rightarrow$

$$\lambda_{1,2} = -s \omega_n \pm i \omega_d$$

$$\omega_n = \sqrt{\frac{g_2}{L}}, \quad s = \frac{d}{2} \sqrt{\frac{L}{g_2}}$$

$$\text{eig}(A, \text{eye}(2,2))$$

$s < 1$, but...

5) deduce stability

linear stability

$$u(t) = \sum_{j=1}^n c_j \lambda_j^j e^{\text{Re}(\lambda_j)t} e^{i \text{Im}(\lambda_j)t}$$

$$\max_{1 \leq j \leq n} \text{Re}(\lambda_j) > 0 \quad \text{UNSTABLE}$$

$$\max_{1 \leq j \leq n} \text{Re}(\lambda_j) < 0 \quad \text{STABLE (linearly)}$$

$$\max_{1 \leq j \leq n} \text{Re}(\lambda_j) = 0 \quad \text{MARGINAL, NEUTRAL (further analysis required)}$$

Diagonalization

2x2 case

$$A\mathcal{X} = \lambda\mathcal{X}$$

$$(\lambda_1, \mathcal{X}^1), (\lambda_2, \mathcal{X}^2) \quad \mathcal{X}^1, \mathcal{X}^2 \text{ linearly independent}$$

Let $S = \begin{pmatrix} \mathcal{X}^1 & \mathcal{X}^2 \\ \text{col 1} & \text{col 2} \end{pmatrix}$ S^{-1} exists ;

then

$$\begin{aligned} AS &= \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \begin{pmatrix} \mathcal{X}_1^1 & \mathcal{X}_1^2 \\ \mathcal{X}_2^1 & \mathcal{X}_2^2 \end{pmatrix} = \begin{pmatrix} A_{11}\mathcal{X}_1^1 + A_{12}\mathcal{X}_2^1 & A_{11}\mathcal{X}_1^2 + A_{12}\mathcal{X}_2^2 \\ A_{21}\mathcal{X}_1^1 + A_{22}\mathcal{X}_2^1 & A_{21}\mathcal{X}_1^2 + A_{22}\mathcal{X}_2^2 \end{pmatrix} \\ &= (A\mathcal{X}^1 \quad A\mathcal{X}^2) = \begin{pmatrix} \lambda_1\mathcal{X}_1^1 & \lambda_2\mathcal{X}_1^2 \\ \lambda_1\mathcal{X}_2^1 & \lambda_2\mathcal{X}_2^2 \end{pmatrix} = \begin{pmatrix} \lambda_1\mathcal{X}_1^1 & & \\ & \lambda_2\mathcal{X}_2^2 & \\ & & \ddots \end{pmatrix} \end{aligned}$$

\mathcal{X}^1
 \mathcal{X}^2

$$\begin{aligned} &= \begin{pmatrix} \mathcal{X}_1^1 & \mathcal{X}_1^2 \\ \mathcal{X}_2^1 & \mathcal{X}_2^2 \end{pmatrix} \begin{pmatrix} \lambda_1 & \\ 0 & \lambda_2 \end{pmatrix} \\ &= S \Lambda \quad \Lambda = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \end{aligned}$$

Hence

$$\begin{aligned} AS &= S\Lambda \Rightarrow A = S\Lambda S^{-1} \\ &\text{or} \\ \Lambda &= S^{-1}AS \end{aligned}$$

diagonalization

nxn case

$$A\mathcal{X} = \lambda\mathcal{X}$$

$$(\lambda_j, \mathcal{X}^j), 1 \leq j \leq n \quad \mathcal{X}_j \text{ linearly independent (assume)}$$

Let

$$S = \begin{pmatrix} \mathcal{X}^1 & \mathcal{X}^2 & \dots & \mathcal{X}^n \\ \text{col 1} & \text{col 2} & \dots & \text{col n} \end{pmatrix}, \quad \Lambda = \begin{pmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \ddots \\ & & & \lambda_n \end{pmatrix};$$

then

$$A = S\Lambda S^{-1}, \quad \Lambda = S^{-1}AS.$$

$$[S, \Lambda] = \text{eig}(A)$$

modal interpretation of ODE NP

Proceed from

$$\frac{dw}{dt} = Aw + F$$

$$\hookrightarrow \frac{dw}{dt} = S\Lambda S^{-1}w + F$$

$$\hookrightarrow \frac{d}{dt} S^{-1}w = \Lambda S^{-1}w + S^{-1}F$$

$$\hookrightarrow \frac{dz}{dt} = \Lambda z + G, \text{ or}$$

$$\frac{dz}{dt} = \begin{pmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \ddots \\ & & & \lambda_n \end{pmatrix} \begin{pmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{pmatrix} + \begin{pmatrix} G_1 \\ G_2 \\ \vdots \\ G_n \end{pmatrix}, \text{ or}$$

$$\begin{cases} S^{-1}w = z \\ \text{new state} \\ \text{variables} \\ S^{-1}F = G \end{cases}$$

$$\frac{dz}{dt} = \begin{pmatrix} \lambda_1 z_1 \\ \lambda_2 z_2 \\ \vdots \\ \lambda_n z_n \end{pmatrix} + \begin{pmatrix} G_1 \\ G_2 \\ \vdots \\ G_n \end{pmatrix}$$

$$\frac{dz_j}{dt} = \lambda_j z_j + G_j, \quad z_j(0) = (S^{-1}w_0)_j \quad 1 \leq j \leq n$$

system of n decoupled ODEs.

Note: for $G_j = 0$,

$$z_j(t) = z_j(0) e^{\operatorname{Re}(\lambda_j)t} e^{i \operatorname{Im}(\lambda_j)t}, \quad 1 \leq j \leq n$$

Many applications of "modal" decomposition...

Computation of Extreme Eigenvalues

Simple power iterations

SPD matrices $B \in \mathbb{R}^{n \times n}$

$$Bx = \lambda x$$

definition (conditions):

- 1) B is symmetric Symmetric
($\Rightarrow \lambda_j$ real, X^j orthogonal)
- 2) λ_j positive Positive Definite
(other equivalent criteria: $v^T B v > 0, v \neq 0; \dots$)

an example: undamped string in tension

Recall ($c_1 = c_2 = 0$; unforced)

$$m' \ddot{u}_i + \frac{T}{h^2} (-u_{i-1} + 2u_i - u_{i+1}) = 0 \quad 1 \leq i \leq n_m$$

zero for $i=1$
zero for $i=n_m$

or

$$\underbrace{\frac{T}{h^2 m'}}_{K_m \text{ (SPD)}} \underbrace{\begin{pmatrix} 2 & -1 & & 0 \\ -1 & 2 & -1 & \\ & & \ddots & \\ 0 & -1 & 2 & -1 \\ & & & -1 & 2 \end{pmatrix}}_u \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_{n_m} \end{pmatrix} = - \underbrace{\begin{pmatrix} \ddot{u}_1 \\ \ddot{u}_2 \\ \vdots \\ \ddot{u}_{n_m} \end{pmatrix}}_{-\ddot{u}}$$

We know (from before) $u = \sum_j c_j \gamma_j e^{i\omega_j t}$

$$\Rightarrow K_m \gamma e^{i\omega t} = -\omega^2 \gamma e^{i\omega t}$$

hence eigenvalues of

$$K_m \gamma = \sigma \gamma \quad \sigma > 0$$

yield frequencies $\omega = \pm \sqrt{\sigma}$.

Typically of (most) interest:

$$\underbrace{\sigma_1}_{\sigma_{\min}} \text{ (smallest eigenvalue)} \Rightarrow \text{lowest frequency.}$$

Power Iteration for λ_{\max} of $Bx = \lambda x$ $B \in \mathbb{R}^{n \times n}$

Recall

$$B = S \Lambda S^{-1} \quad S = \begin{pmatrix} x^1 & x^2 & \dots & x^n \end{pmatrix}, \quad \Lambda = \begin{pmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{pmatrix}$$

$0 < \lambda_1 < \lambda_2 < \dots < \lambda_n$
 $\lambda_{\min} \qquad \qquad \qquad \lambda_{\max}$

Thus

$$B^k = \underbrace{(S \Lambda S^{-1}) (S \Lambda S^{-1}) \dots (S \Lambda S^{-1})}_{k \text{ terms}}$$

$$= S \Lambda^k S^{-1}$$

$$= S \underbrace{\begin{pmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{pmatrix} \begin{pmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{pmatrix} \dots \begin{pmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{pmatrix}}_{k \text{ terms}} S^{-1}$$

$$= S \begin{pmatrix} \lambda_1^k & & & \\ & \lambda_2^k & & \\ & & \ddots & \\ & & & \lambda_n^k \end{pmatrix} S^{-1}$$

so

$$B^k x_0 = \begin{pmatrix} x^1 & x^2 & \dots & x^n \end{pmatrix} \begin{pmatrix} \lambda_1^k & & & \\ & \lambda_2^k & & \\ & & \ddots & \\ & & & \lambda_n^k \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix}$$

$\underbrace{S^{-1} x_0}_{\begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix}}$

$$= \begin{pmatrix} x^1 & x^2 & \dots & x^n \end{pmatrix} \begin{pmatrix} \alpha_1 \lambda_1^k \\ \alpha_2 \lambda_2^k \\ \vdots \\ \alpha_n \lambda_n^k \end{pmatrix}$$

$$= \alpha_1 \lambda_1^k x^1 + \alpha_2 \lambda_2^k x^2 + \dots + \alpha_n \lambda_n^k x^n$$

$$\rightarrow \alpha_n \lambda_n^k x^n \text{ as } k \rightarrow \infty.$$

algorithm

given $x^{(0)}$, $k = 1$:

for $k = 1 : k_{max}$
 $v = Bx^{(k-1)}$

$x^{(k)} = v / \|v\|$ ($\Rightarrow \|x^{(k)}\|^2 = \|v\|^2 / \|v\|^2 = 1$) avoid overflow

$\lambda_{max}^{(k)} = (x^{(k)})^T B x^{(k)}$ $\underbrace{\|x^{(k)}\|^2}_1$

end

Note:

$(x^n)^T B x^n = (x^n)^T \lambda_n x^n = \lambda_{max}$

if

$(x^n)^T (x^n) = \|x^n\|^2$ (normalized) to unity.

Inverse

Power Iteration for λ_{min} of $Bx = \lambda x$ $B \in \mathbb{R}^{n \times n}$

Recall $B = S \Lambda S^{-1}$

$\Rightarrow B^{-1} = S \Lambda^{-1} S^{-1}$ $S = (x^1 x^2 \dots x^n)$, $\Lambda = \begin{pmatrix} \lambda_1^{-1} & & \\ & \lambda_2^{-1} & \\ & & \dots \\ & & & \lambda_n^{-1} \end{pmatrix}$
 $0 < \lambda_1 < \lambda_2 < \dots < \lambda_n$
 $\lambda_{min} \qquad \qquad \qquad \lambda_{max}$

Thus

$(B^{-1})^k = \overbrace{(S \Lambda^{-1} S^{-1})(S \Lambda^{-1} S^{-1}) \dots (S \Lambda^{-1} S^{-1})}^{k \text{ terms}}$
 $= S (\Lambda^{-1})^k S^{-1}$
 $= S \begin{pmatrix} \lambda_1^{-k} & & \\ & \lambda_2^{-k} & \\ & & \dots \\ & & & \lambda_n^{-k} \end{pmatrix} \begin{pmatrix} \lambda_1^{-1} & & \\ & \lambda_2^{-1} & \\ & & \dots \\ & & & \lambda_n^{-1} \end{pmatrix} \dots \begin{pmatrix} \lambda_1^{-1} & & \\ & \lambda_2^{-1} & \\ & & \dots \\ & & & \lambda_n^{-1} \end{pmatrix} S^{-1}$

$= S \begin{pmatrix} \lambda_1^{-k} & & \\ & \lambda_2^{-k} & \\ & & \dots \\ & & & \lambda_n^{-k} \end{pmatrix} S^{-1}$

so

$(B^{-1})^k x_0 = \begin{pmatrix} x^1 & x^2 & \dots & x^n \end{pmatrix} \begin{pmatrix} \lambda_1^{-k} & & \\ & \lambda_2^{-k} & \\ & & \dots \\ & & & \lambda_n^{-k} \end{pmatrix} \underbrace{\begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix}}_{S^{-1} x_0}$
 $= \begin{pmatrix} x^1 & x^2 & \dots & x^n \end{pmatrix} \begin{pmatrix} \alpha_1 \lambda_1^{-k} \\ \alpha_2 \lambda_2^{-k} \\ \vdots \\ \alpha_n \lambda_n^{-k} \end{pmatrix}$
 $= \alpha_1 \lambda_1^{-k} x^1 + \alpha_2 \lambda_2^{-k} x^2 + \dots + \alpha_n \lambda_n^{-k} x^n$
 $\rightarrow \alpha_1 \lambda_1^{-k} x_1$ as $k \rightarrow \infty$.

algorithm 1

given $x^{(0)}$:

for $k = 1 : k_{max}$

$Bv = x^{(k-1)}$ ($v = B^{-1} x^{(k-1)}$)

$x^{(k)} = v / \|v\|$ ($\Rightarrow \|x^{(k)}\|^2 = 1$)

$\lambda^{(k)} = x^{(k)T} B x^{(k)}$

end

Note:

$(x_1)^T B x_1 = (x_1)^T \lambda_1 x_1 = \lambda_{min}$

if

$(x_1)^T (x_1) = \|x_1\|^2$ (normalized) to unity.

algorithm 2

more efficient

given \hat{x} , $\hat{\lambda}$, ϵ :
while $\|B\hat{x} - \hat{\lambda}\hat{x}\| > \epsilon$
 $Bv = \hat{x}$
 $\beta = \|v\|$
 $\hat{\lambda} = v^T \hat{x} / \beta^2$
 $\hat{x} = v / \beta$
end

(Improvements: shift, ...)

MIT OpenCourseWare
<http://ocw.mit.edu>

2.086 Numerical Computation for Mechanical Engineers
Spring 2013

For information about citing these materials or our Terms of Use, visit: <http://ocw.mit.edu/terms>.