2.092/2.093 — Finite Element Analysis of Solids & Fluids I

Fall '09

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Lecture 21 - Solution of the Generalized Eigenvalue Problem

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Reading assignment: Chapters 10 and 11

$$M\ddot{U} + KU = R \tag{1}$$

Aside: M could have zero masses. Then we use Gauss elimination on K to remove zero-mass DOFs, but we denote the final matrix still as K. Then, in free vibrations:

$$M\ddot{U} + KU = 0 \tag{2}$$

where now M and K are assumed to be positive definite matrices, i.e.  $\tilde{U}^T M \tilde{U} > 0$ ,  $\tilde{U}^T K \tilde{U} > 0$  for any  $U \neq 0$ . Then, we obtain the eigenvalue problem

$$K\phi = \lambda M\phi \quad \rightarrow \quad K\phi_i = \lambda_i M\phi_i$$
 (A)

where  $0 < \underbrace{\lambda_1}{\phi_1} \leq \underbrace{\lambda_2}{\phi_2} \leq \ldots \leq \underbrace{\lambda_n}{\phi_n}$ .

Recall:

$$egin{aligned} oldsymbol{\phi}_i^T oldsymbol{M} oldsymbol{\phi}_j &= \delta_{ij} \ oldsymbol{\phi}_i^T oldsymbol{K} oldsymbol{\phi}_j &= \omega_i^2 \delta_{ij} = \lambda_i \delta_{ij} \end{aligned}$$

## The Case of Multiple Eigenvalues

Assume  $\lambda_1 = \lambda_2 < \lambda_3$ , i.e.  $\lambda_1$  has a multiplicity of 2 (m = 2),  $\phi_1$  and  $\phi_2$  are two eigenvectors for  $\lambda_1$  and  $\lambda_2$ , and  $\phi_1 \neq \phi_2$ . Then, we have

$$\boldsymbol{K}\alpha\boldsymbol{\phi}_1 = \lambda_1 \boldsymbol{M}\alpha\boldsymbol{\phi}_1 \quad (\alpha: \text{ any constant}) \tag{3}$$

$$\boldsymbol{K}\boldsymbol{\beta}\boldsymbol{\phi}_2 = \lambda_1 \boldsymbol{M}\boldsymbol{\beta}\boldsymbol{\phi}_2 \quad (\boldsymbol{\beta}: \text{ any constant})$$
(4)

Hence,

$$\boldsymbol{K}\left(\alpha\boldsymbol{\phi}_{1}+\boldsymbol{\beta}\boldsymbol{\phi}_{2}\right)=\lambda_{1}\boldsymbol{M}\left(\alpha\boldsymbol{\phi}_{1}+\boldsymbol{\beta}\boldsymbol{\phi}_{2}\right)$$
(5)

Eq. (5) shows  $\alpha \phi_1 + \beta \phi_2 = \tilde{\phi}$  is also an eigenvector corresponding to  $\lambda_1$ ! We can change the length of the eigenvector so that for some  $\gamma$ ,

$$\left(\gamma\tilde{\pmb{\phi}}
ight)^{T}\pmb{M}\left(\gamma\tilde{\pmb{\phi}}
ight)=1$$

Recall we want  $\ddot{x}_i + \omega_i^2 x_i = r_i$ , having set the mass m to 1 since  $\phi_i^T M \phi_i = \delta_{ij}$ .

If the eigenvalues for the system (A) are distinct, the eigenvectors are unique. Here, we have a two dimensional eigenspace  $(\lambda_1 = \lambda_2)$ . Any two *M*-orthogonal vectors in this space are eigenvectors and could be used as mode shapes.

## Gram-Schmidt (see textbook)

Orthogonalization is used to obtain M-orthogonal vectors. For an eigenvalue of multiplicity m, we have an eigenspace of dimension m and can always find m *M*-orthogonal vectors that are in this eigenspace. We need orthogonality to decouple Eq. (2). Next, we will discuss some solution techniques.

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## **Inverse Iteration**

Once we have eigenvectors with  $\phi_i^T M \phi_j = \delta_{ij}$ , we could simply use  $\phi_i^T K \phi_j = \lambda_i \delta_{ij}$  to obtain  $\lambda_i$ .

Do we need to iterate on  $\mathbf{K}\boldsymbol{\phi} = \lambda(\mathbf{M}\boldsymbol{\phi})$  to get  $\mathbf{K}\boldsymbol{\phi}_i = \lambda_i \mathbf{M}\boldsymbol{\phi}_i$ ? Since for the general case there are no explicit formulas available to calculate the roots of  $p(\lambda)$  when the order of p is greater than 4, an iterative solution method has to be used.

## Iteration

Assume  $\lambda_1 > 0$ . We pick  $\boldsymbol{x}_1$  and use for  $k = 1, 2, \ldots$ 

$$egin{aligned} & Kar{x}_{k+1} = Mx_k \ & (a) \ & x_{k+1} = rac{ar{x}_{k+1}}{ig(ar{x}_{k+1}^T Mar{x}_{k+1}ig)^rac{1}{2}} \end{aligned}$$

Since  $\lambda_1 > 0$ , K is positive definite and we can solve Eq. (a). We want  $x_{k+1}$  to satisfy the mass orthonormality relation  $\bar{x}_{k+1}^T M \bar{x}_{k+1} = 1$ . If we assume  $x_1^T M \phi_1 \neq 0$ , then

$$\boldsymbol{x}_{k+1} \rightarrow \boldsymbol{\phi}_1 \text{ as } k \rightarrow \infty$$
  
 $\lambda_1 = \boldsymbol{\phi}_1^T \boldsymbol{K} \boldsymbol{\phi}_1 , \, \boldsymbol{\phi}_1^T \boldsymbol{M} \boldsymbol{\phi}_1 = 1$   
 $\boldsymbol{K} \boldsymbol{x}_{k+1} = \boldsymbol{M} \boldsymbol{x}_k$  (B)

Proof: Consider

We see that (B) is equivalent to working with vectors  $z_{k+1}$  and  $z_k$ .

$$\mathbf{\Phi} oldsymbol{z}_{k+1} = oldsymbol{x}_{k+1}$$
 ,  $\mathbf{\Phi} oldsymbol{z}_k = oldsymbol{x}_k$ 

Substitute into (B):

$$\Phi^{T} \boldsymbol{K} \Phi \boldsymbol{z}_{k+1} = \Phi^{T} \boldsymbol{M} \Phi \boldsymbol{z}_{k}$$

$$\begin{bmatrix} \lambda_{1} & \text{zeros} \\ \lambda_{2} & \\ & \ddots & \\ \text{zeros} & \lambda_{n} \end{bmatrix} \boldsymbol{z}_{k+1} = \boldsymbol{z}_{k}$$
(C)

. 1

Working on (C) is equivalent to working on (B)

m

Next, iterate with (C). Assume:

Then we find

After l iterations,

$$oldsymbol{z}_2^T = \left[ egin{array}{cccccc} rac{1}{\lambda_1} & rac{1}{\lambda_2} & rac{1}{\lambda_3} & \cdots & rac{1}{\lambda_n} \end{array} 
ight]$$

$$oldsymbol{z}_{l+1}^T = \left[ egin{array}{c} \left(rac{1}{\lambda_1}
ight)^l & \left(rac{1}{\lambda_2}
ight)^l & \left(rac{1}{\lambda_3}
ight)^l & \ldots & \left(rac{1}{\lambda_n}
ight)^l \end{array} 
ight]$$

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Only the direction of the vector is important.

Assume  $\lambda_1 < \lambda_2$ . Multiply  $\boldsymbol{z}_{l+1}$  by  $(\lambda_1)^l$  to obtain a new  $z_{l+1}$ :

$$\boldsymbol{z}_{l+1}^T = \left[ \begin{array}{ccc} 1 & \left(\frac{\lambda_1}{\lambda_2}\right)^l & \left(\frac{\lambda_1}{\lambda_3}\right)^l & \dots & \left(\frac{\lambda_1}{\lambda_n}\right)^l \end{array} 
ight]$$

This  $\boldsymbol{z}_{l+1}^T$  converges to  $\begin{bmatrix} 1 & 0 & 0 & \dots & 0 \end{bmatrix}$  as  $l \to \infty$ . Note that if  $\boldsymbol{z}_1$  is orthogonal to  $\begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$ , we will never reach the eigenvector corresponding to  $\lambda_1$ .

Finally, assume  $\lambda_1 = \lambda_2 < \lambda_3$ . Then we obtain

$$\boldsymbol{z}_{l+1}^T = \begin{bmatrix} 1 & 1 & 0 & \dots & 0 \end{bmatrix}$$

To obtain the 2nd eigenvector for  $\lambda_1 = \lambda_2$ , choose a starting vector  $x_1$  that is *M*-orthogonal to  $\phi_1$  and enforce this orthogonality in each iteration. To avoid round-off error, see the textbook.

In practice, the inverse iteration method is hardly used by itself, but rather as an ingredient in a more complex scheme. The next lecture introduces the widely used "subspace iteration method" which employs the inverse iteration method to efficiently solve for the first few lowest frequencies/eigenvalues and modeshapes of large systems.

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