

Method of Finite Elements II

Example 11.4, 11.5 & 11.8
and necessary theory

Fabio Wider
Centre for Mechanics

$$\mathbf{K}\Phi = \lambda\mathbf{M}\Phi$$

Four groups of solution methods:

- Vector Iteration

$$\mathbf{K}\Phi_i = \lambda_i\mathbf{M}\Phi_i$$

- Transformation methods
- Polynomial iteration techniques
- Using the Sturm sequence property of characteristic polynomials

We usually use as a start vector $\mathbf{x}_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$

We assume that \mathbf{K} is pd and $\mathbf{y}_1 = \mathbf{M}\mathbf{x}_1$.

And by evaluating for $k = 1, 2, \dots$

$$\mathbf{K}\bar{\mathbf{x}}_{k+1} = \mathbf{y}_k$$

$$\bar{\mathbf{y}}_{k+1} = \mathbf{M}\bar{\mathbf{x}}_{k+1}$$

$$\rho(\bar{\mathbf{x}}_{k+1}) = \frac{\bar{\mathbf{x}}_{k+1}^T \mathbf{y}_k}{\bar{\mathbf{x}}_{k+1}^T \bar{\mathbf{y}}_{k+1}}$$

$$\mathbf{y}_{k+1} = \frac{\bar{\mathbf{y}}_{k+1}}{\left(\bar{\mathbf{x}}_{k+1}^T \bar{\mathbf{y}}_{k+1}\right)^{1/2}}$$

Provided that $\mathbf{y}_1^T \mathbf{\Phi}_1 \neq 0$

$$\mathbf{y}_{k+1} \rightarrow \mathbf{M}\mathbf{\Phi}_1 \text{ and } \rho(\bar{\mathbf{x}}_{k+1}) \rightarrow \lambda_1 \text{ as } k \rightarrow \infty$$

If l is the last iteration, we end up having $\lambda_1 = \rho(\bar{\mathbf{x}}_{l+1})$

$$\mathbf{\Phi}_1 = \frac{\bar{\mathbf{x}}_{l+1}}{\left(\bar{\mathbf{x}}_{l+1}^T \bar{\mathbf{y}}_{l+1}\right)^{1/2}}$$

... the first eigenpair!

Only thing that changes: we
assume \mathbf{M} to be pd and

$$\mathbf{y}_1 = \mathbf{K}\mathbf{x}_1$$

$$\mathbf{M}\bar{\mathbf{x}}_{k+1} = \mathbf{y}_k$$

$$\bar{\mathbf{y}}_{k+1} = \mathbf{K}\bar{\mathbf{x}}_{k+1}$$

$$\rho(\bar{\mathbf{x}}_{k+1}) = \frac{\bar{\mathbf{x}}_{k+1}^T \bar{\mathbf{y}}_k}{\bar{\mathbf{x}}_{k+1}^T \mathbf{y}_{k+1}}$$

$$\mathbf{y}_{k+1} = \frac{\bar{\mathbf{y}}_{k+1}}{\left(\bar{\mathbf{x}}_{k+1}^T \mathbf{y}_{k+1}\right)^{1/2}}$$

Provided that $\Phi_n^T \mathbf{y}_1 \neq 0$

$$\mathbf{y}_{k+1} \rightarrow \mathbf{K}\Phi_n \text{ and } \rho(\bar{\mathbf{x}}_{k+1}) \rightarrow \lambda_n \text{ as } k \rightarrow \infty$$

and if l is again the last iteration, we have

$$\lambda_n = \rho(\bar{\mathbf{x}}_{l+1})$$

$$\Phi_n = \frac{\bar{\mathbf{x}}_{l+1}}{\left(\bar{\mathbf{x}}_{l+1}^T \mathbf{y}_l\right)^{1/2}}$$

...the last eigenpair!

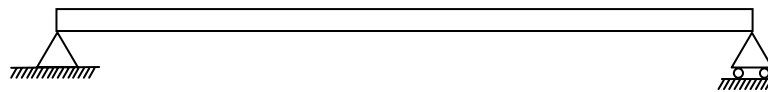
In both iteration procedures the convergence is measured by

$$\frac{|\lambda_l^{(k+1)} - \lambda_l^{(k)}|}{\lambda_l^{(k+1)}} \leq tol$$

Where *tol* is usually some decimals, e.g. 1e-8

Use forward iteration with $tol = 1e-6$ to evaluate λ_4 and Φ_4 of the eigenproblem $\mathbf{K}\Phi = \lambda\mathbf{M}\Phi$, where

$$\mathbf{K} = \begin{pmatrix} 5 & -4 & 1 & 0 \\ -4 & 6 & -4 & 1 \\ 1 & -4 & 6 & -4 \\ 0 & 1 & -4 & 5 \end{pmatrix}; \mathbf{M} = \begin{pmatrix} 2 & & & \\ & 2 & & \\ & & 1 & \\ & & & 1 \end{pmatrix}$$



Using the starting vector $\mathbf{x}_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$

Using MATLAB, we get the following fourth eigenpair...

k	$\bar{\mathbf{x}}_{k+1}$	$\bar{\mathbf{y}}_{k+1}$	$\rho(\bar{\mathbf{x}}_{k+1})$	\mathbf{y}_{k+1}	$\frac{ \lambda_4^{(k+1)} - \lambda_4^{(k)} }{\lambda_4^{(k+1)}}$
10	-1.1416 2.7170 -7.7476 5.9816	-24.3237 57.8405 -82.4219 63.6157	10.63845	-2.2864 5.4369 -7.7476 5.9798	0.0000009437

$$\lambda_4 \doteq 10.63845; \quad \boldsymbol{\phi}_4 \doteq \begin{bmatrix} -0.10731 \\ 0.25539 \\ -0.72827 \\ 0.56227 \end{bmatrix}$$

Improving convergence rate?

What when \mathbf{M} and \mathbf{K} are not pd?

Applying a shift μ :

$$(\mathbf{K} - \mu\mathbf{M})\Phi = \eta\mathbf{M}\Phi$$

Use inverse iteration in order to calculate (λ_1, Φ_1) of the problem $\mathbf{K}\Phi = \lambda\mathbf{M}\Phi$, where \mathbf{K} and \mathbf{M} are given in Example 11.4. Then impose a shift $\mu = 10$ and show that in the inverse iteration convergence occurs toward (λ_4, Φ_4) . Use again a tolerance of $1e-6$.

After three iterations we get for the first eigenpair...

$$\lambda_1 \doteq 0.09654; \quad \phi_1 \doteq \begin{bmatrix} 0.3126 \\ 0.4955 \\ 0.4791 \\ 0.2898 \end{bmatrix}$$

Then imposing a shift of $\mu = 10$, we obtain...

$$\mathbf{K} - \mu\mathbf{M} = \begin{pmatrix} -15 & -4 & 1 & 0 \\ -4 & -14 & -4 & 1 \\ 1 & -4 & -4 & -4 \\ 0 & 1 & -4 & -5 \end{pmatrix}$$

Using now again the inverse iteration on the problem $(\mathbf{K} - \mu\mathbf{M})\Phi = \eta\mathbf{M}\Phi$
we obtain convergence after six iterations and get...

$$\rho(\bar{\mathbf{x}}_7) = 0.6385; \quad \mathbf{x}_7 = \begin{bmatrix} -0.1076 \\ 0.2556 \\ -0.7283 \\ 0.5620 \end{bmatrix}$$

Since we imposed a shift, we know:

- $\mu + \rho(\bar{\mathbf{x}}_7)$ is approximation to an eigenvalue
- \mathbf{x}_7 is approximation to the corresponding eigenvector

We do not know:

- which eigenpair is approximated

Solution here: by comparing with Example 11.4, we see that we have the fourth eigenpair.

In case that no other example is available: Do not choose shift arbitrarily, use e.g. Rayleigh Quotient Iteration. (Bathe 11.2.4)

Orthogonalize starting vector to eigenvectors already calculated

→ Convergence occurs only to other eigenvectors

Suppose:

- Calculated are eigenvectors $\Phi_1, \Phi_2, \dots, \Phi_m$
- Now \mathbf{M} -orthogonalize \mathbf{x}_1 to these eigenvectors (deflate \mathbf{x}_1)

$$\tilde{\mathbf{x}}_1 = \mathbf{x}_1 - \sum_{i=1}^m \alpha_i \Phi_i \quad \text{where} \quad \alpha_i = \Phi_i^T \mathbf{M} \mathbf{x}_1; \quad i = 1, \dots, m$$

Gram-Schmidt Orthogonalization

In inverse iteration $\tilde{\mathbf{x}}_1$ is now used as a starting vector instead of \mathbf{x}_1 .

Convergence occurs now to eigenpair $(\lambda_{m+1}, \Phi_{m+1})$, which is unknown yet.

Calculate, using Gram-Schmidt orthogonalization, an appropriate starting iteration vector for the solution of the problem $\mathbf{K}\Phi = \lambda\mathbf{M}\Phi$, where \mathbf{K} and \mathbf{M} are given in Example 11.4. Assume that the eigenpairs (λ_1, Φ_1) and (λ_4, Φ_4) , are known as obtained in Example 11.5 and that convergence to another eigenpair is sought.

Deflate unit full vector of the known eigenvectors:

$$\tilde{\mathbf{x}}_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} - \alpha_1 \mathbf{\Phi}_1 - \alpha_4 \mathbf{\Phi}_4$$

where the α 's are obtained by $\alpha_1 = \mathbf{\Phi}_1^T \mathbf{M} \mathbf{x}_1$; $\alpha_4 = \mathbf{\Phi}_4^T \mathbf{M} \mathbf{x}_1$

Substituting the values obtained in Example 11.5 leads us to...

$$\alpha_1 = 2.385; \alpha_2 = 0.1299$$

So we finally end up having...

$$\tilde{\mathbf{x}}_1 = \begin{pmatrix} 0.2683 \\ -0.2149 \\ -0.04812 \\ 0.2358 \end{pmatrix}$$

Thank you for your attention.