Higher Order Eigenpair Perturbations

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A matrix method is presented for the exact calculation of large eigenpair perturbations arising from design variable changes. The method is "forward analysis" in the sense that design variable changes are known and eigenpair perturbations are computed. Competing methods for forward analysis include reanalysis and first-order approximate methods. First-order methods are well known for their ability to calculate small eigenpair changes but are inadequate when large changes are required. In this paper, a new method of forward analysis is presented that takes into account all orders of the perturbation expansion. A benchmark cantilever beam will be analyzed with the new method for large perturbations in design variables. The method will be valuable for fast reanalysis and for identification and model correlation studies where sizeable differences exist between a baseline model and an objective model.

Nomenclature

\( c_i \) = weighting factor determining the contribution of the homogeneous solution to the total solution for \( \Delta \phi_i \)

\( [D]_i \) = coefficient matrix in solution for \( \Delta \phi_i \); equals \( [K] - \lambda_i [M] \)

\( [F]^1_i \) = static pseudoload vector appearing in linearized \( (\Delta \phi_i) \) calculation, \( = (\Delta \lambda_i[M] + \lambda_i^2 [M]) \)

\( -[K] (\phi_i) \); orthogonal to \( \{ \phi_i \} \)

\( [F]^n_i \) = static pseudoload vector appearing in nonlinear \( (\Delta \phi_i) \) calculation, \( = (\Delta \lambda_i[M]^n + \lambda_i^n [M]) \)

\( -[K] (\phi_i) \); orthogonal to \( \{ \phi_i \} \)

\( [K] \) = stiffness matrix, symmetric

\( [M] \) = mass matrix, symmetric

\( \Delta \) = perturbation symbol denoting exact change from the baseline

\( \lambda_i \) = \( i \)th eigenvalue (circular frequency squared)

\( \{ \phi_i \} \) = \( i \)th eigenvector (i.e., mode shape of structure); solution to \( [K] (\phi_i) = \lambda_i [M] (\phi_i) \); normalized with respect to appropriate \( [M] \)

Superscripts

0 = baseline values
1 = perturbed values resultant from design variable change, e.g., \( (\phi)^1 = (\phi)^0 + \Delta(\phi) \)

Introduction

CURREN T methods for optimization in structural dynamics use derivatives with respect to design variables. When numerically calculating these derivatives, one commonly uses perturbations of 2\% or less in the design variables. With design changes this small, the eigenpair perturbations calculated are small, and the resulting sensitivities are derivatives at the baseline configuration. Currently popular methods of calculating eigenpair sensitivities all compute the derivative values, either directly through differentiating the eigenproblem statement or indirectly through calculating small change in first-order perturbation equations (having dropped higher order terms). It is important to realize that the perturbation equations of first order in \( \Delta \) are the same as the derivative equations.

Literature Survey and Research Impetus

Fox and Kapoor's method differentiates the eigenproblem equation and calculates derivatives by a modal superposition method. It usually requires a full set of modes to obtain exact derivatives (counterexample: tip mass as design variable in Ref. 2). Wang's method modifies Fox and Kapoor's method for the case of a truncated set of modes through the use of a residual static mode. Nelson's method calculates derivatives through removing the singularity in the matrix equation and solving directly. It needs information associated with the \( i \)th mode only, so the issue of using a truncated modal set is not important. High's method iteratively computes small eigenpair perturbations and, for large systems, is faster than direct calculation. This method has been implemented in MSC/NASTRAN, but the iterations are divergent for anything other than small change. Surveys and comparisons of these and other small change methods are readily available in the literature.

The work presented here departs from small change ideas and calculates large changes in the eigenpairs by handling the nonlinear nature of the perturbation equations. Kim et al., and Hoff et al. have explored similar directions. Another recent paper developed independently by Kim is very similar in spirit to the current work, but the solution procedure is different and a first-order normalization is used. As presented, this paper's method is most easily characterized as fast reanalysis, since the changes calculated are exact. If large changes are present, dividing the eigenpair perturbations by the change in the design variable yields secant sensitivity values instead of tangent derivative values. Thus, the proper term might be eigenpair secant sensitivities rather than eigenpair derivatives.

The retention of all terms in the expansion of the governing equations (eigenproblem, normalization) is required to study large changes in the eigenpairs resulting from large changes in design variables. Consideration of large change is unnecessary in the current applications of eigenpair derivative technology.
known to the authors. Potential use can be foreseen in the areas of model correlation and system identification in which large exact changes would be more desirable than small approximate changes. Although these are generally "inverse analysis" eigenpair perturbations known, design variable perturbations desired, the forward analysis would be necessary in a predictor/corrector algorithm. As an example of large change, consider damage to a large space truss in which an element has been destroyed. This requires a 100% reduction in stiffness for that element (a redundant structure is required). The small change methods referenced previously would fail in calculating meaningful eigenpair derivatives since they are generally restricted to changes of only a few percent. Another potential use is optimization work in which criteria other than frequencies and mode shapes drive the design and in which frequencies and mode shapes are allowed to drift (or are inactive constraints). When this is the case, large changes in frequencies and modes may occur on each redesign iteration, and large change algorithms would be needed to track them.

Repeated Eigenvalues

The subject of repeated eigenvalues has recently received a great deal of attention. Modal superposition techniques such as Fox and Kapoor's method are particularly susceptible to problems when multiplicity is present. The work of Kim and Wallerstein is most easily applied to this paper's technology. They employ an orthogonal transformation to generate a set of complementary mode shapes. All baseline mode shapes are required. The only modification necessary for the authors' method to handle repeated eigenvalues, according to Kim, is the use of these complementary mode shapes in place of the baseline mode shapes.

Equation Derivations

Turning attention to the derivation of the governing equations, consider the classical, linear, structural eigenproblem:

$$[K^0][\phi_0] = \lambda_0^0[M^0][\phi_0],$$

(1)

When the system is perturbed from the baseline through mass and stiffness changes, the new statement of dynamic equilibrium is

$$[K^i][\phi_i] = \lambda_i^i[M^i][\phi_i],$$

(2)

or in expanded form

$$(K^0) + (\Delta K)(\phi_i) + (\Delta \phi_i)$$

$$= (\lambda_0^i + \Delta \lambda_i)(M^i + \Delta M)(\phi_i) + (\Delta \phi_i).$$

(3)

Expanding, canceling the baseline solution, and regrouping yields

$$([K^0] - \lambda_0^i[M^0])(\Delta \phi_i) + ([\Delta K] - \lambda_0^i[\Delta M])(\phi_i) + (\Delta \phi_i)$$

$$= \Delta \lambda_i([M^0] + [\Delta M])(\phi_i) + (\Delta \phi_i).$$

(4)

Premultiplying by $[\phi_i]^T$ and canceling the baseline solution again due to the symmetry of $[M^0]$ and $[K^0]$ yields

$$\lambda^0_i([K^0] - \lambda_0^i[M^0])(\phi_i) + (\Delta \phi_i)$$

$$= \Delta \lambda_i[\phi_i]^T([M^0] + [\Delta M])(\phi_i) + (\Delta \phi_i).$$

(5)

First-Order System of Equations

Popular derivative solutions and first-order approximations (specifically Fox and Kapoor's method, Wang's method, Nelson's method, and High's method) all solve equations derivable by dropping the terms of higher order in $\Delta$. When this is done, Eq. (5) becomes

$$[\phi_i]^T([\Delta K] - \lambda_0^i[\Delta M])(\phi_i) = \Delta \lambda_i[\phi_i]^T[M^0][\phi_i].$$

(6)

Employing the normalization condition $[\phi_i]^T[M^0][\phi_i] = 1$, one obtains the following uncoupled equation for $\Delta \lambda_i$:

$$\Delta \lambda_i = [\phi_i]^T([\Delta K] - \lambda_0^i[\Delta M])(\phi_i).$$

(7)

In a like manner, one drops the higher order terms from Eq. (4) and solves for $[\Delta \phi_i]$ to obtain the other important first-order formula:

$$([K^0] - \lambda_0^i[M^0])(\Delta \phi_i) = (\Delta \lambda_i[M^0] + \lambda_0^i[\Delta M] - [\Delta K])(\phi_i).$$

(8)

This formula can be written as

$$[D^0][\Delta \phi_i] = \{F^i\},$$

(9)

where

$$[D^0] = [K^0] - \lambda_0^i[M^0]$$

(10)

is singular and

$$\{F^i\} = (\Delta \lambda_i[M^0] + \lambda_0^i[\Delta M] - [\Delta K])(\phi_i).$$

(11)

This can be viewed as a static pseudoload. This equation corresponds to the pathological Fredholm alternative in which the coefficient matrix is singular and there is a nonzero load. Such equations cannot be solved in general. This equation is solvable, however, since it is "consistent," i.e., $\{F^i\}$ is orthogonal to $[\phi_i]$. (Note: the fact that this first-order system is consistent must be viewed as fortunate.)

To prove the consistency of the linearized problem, one premultiplies $\{F^i\}$ by the transpose of the baseline eigenvector:

$$[\phi_i]^T[F^i_i] = \Delta \lambda_i(\phi_i)^T[M^0][\phi_i] + \lambda_0^i[\phi_i]^T[\Delta M][\phi_i]$$

$$- [\phi_i]^T[\Delta K][\phi_i].$$

(12)

Substituting the first-order equation for $\Delta \lambda_i$ [Eq. (7)] cancels all terms on the right-hand side:

$$[\phi_i]^T[F^i_i] = [\phi_i]^T[\Delta K][\phi_i] - \lambda_0^i[\phi_i]^T[\Delta M][\phi_i]$$

$$+ \lambda_0^i[\phi_i]^T[\Delta M][\phi_i] - [\phi_i]^T[\Delta K][\phi_i]$$

$$= 0$$

(13)

Nonlinear System of Equations

The focus of this paper will be on the nonlinear perturbation equations and the retention of all higher order terms. The first nonlinear formula is obtained by solving Eq. (5) for $\Delta \lambda_i$:

$$\Delta \lambda_i = [\phi_i]^T([\Delta K] - \lambda_0^i[\Delta M])(\phi_i)$$

$$/[\phi_i]^T[M^0][\phi_i].$$

(14)

The second nonlinear equation is derived by rearranging the terms of Eq. (4) so as to collect all $[\Delta \phi_i]$ terms on the left-hand side:

$$([K^0] - \lambda_0^i[M^0])(\Delta \phi_i) + ([\Delta K] - \lambda_0^i[\Delta M])(\Delta \phi_i)$$

$$= \Delta \lambda_i([M^0] + [\Delta M])(\Delta \phi_i)$$

$$= \Delta \lambda_i([M^0] + [\Delta M])(\Delta \phi_i).$$

(15)
Combining terms yields the relation
\[(K^\prime') - \lambda^2_i [M'] \{\Delta \phi\} = (\Delta \lambda_i [M'] + \lambda^2_i [\Delta M] - [\Delta K]) \{\phi^0\},\]
(16)
This equation, as its linearized counterpart, can be written
\[\{D^\prime\} \{\Delta \phi\} = \{F''\},\]
(17)
where
\[\{D^\prime\} = (K^\prime') - \lambda^2_i [M']\]
(18)
is singular and
\[\{F''\} = (\Delta \lambda_i [M'] + \lambda^2_i [\Delta M] - [\Delta K]) \{\phi^0\},\]
(19)
is a static pseudo-load. This equation is also consistent, though this time the pseudo-load vector is orthogonal to \{\phi^0\}, instead of \{\phi^0\}.

The proof of the consistency for the nonlinear problem can be performed by premultiplying \{F''\} by the transpose of the perturbed eigenvector. The same cancellation of terms occurs as in the linear case after substituting the exact equation for \Delta \lambda_i [Eq. (14)] and employing the symmetry of \{M'\}, \{\Delta M\}, and \{\Delta K\}.

Equations (14) and (17) look quite similar in form to Eqs. (7) and (9). Their solution methodology is very different, however. Equations (14) and (17) are coupled since \Delta \lambda_i is dependent on \{\phi^0\} and \{\Delta \phi\}; \lambda_i is dependent on \{\phi^0\}. Simultaneous solution seems impossible. Thus, solution of this system must be done iteratively. The first-order system, on the other hand, may be solved directly since \Delta \lambda_i depends only on known quantities in Eq. (7).

Theoretical Development
Solution for the exact eigenpair changes is now a problem of correctly solving Eqs. (14) and (17). This is not easy due to the coupling of the equations and to the inherent instability of the problem. The proposed method of attack is to find an initial guess for \Delta \lambda_i and then to iterate Eqs. (14) and (17).

There are four important points in the mathematics of the problem that must be incorporated in the iteration scheme. These points are 1) the switching of solution logic as \{D^\prime\}, is driven singular, 2) the deletion of the spurious solution for \{\Delta \phi\}, 3) the proper management of the homogeneous and particular solutions for \{\Delta \phi\}, and 4) the calculation of the particular solution.

Point 1: Switching of Solution Logic
The singularity of the matrix \{D^\prime\} is very important and is dependent on the accuracy of the \lambda_i estimate at a particular iteration. For a nonexact estimate, the \{D^\prime\} matrix is nonsingular and can be inverted for an estimate of \{\Delta \phi\}. When the \lambda_i estimate has converged, the matrix is singular and what is herein called a modified Nelson's method is used to find the particular solution, which is in turn used to find \{\Delta \phi\}.

On the starting iteration, the \lambda_i estimate is relatively poor and \{D^\prime\} is nonsingular. There are no homogeneous and particular solutions to manage, and \{\Delta \phi\} is found from standard matrix decomposition techniques. This solution for \{\Delta \phi\} will be heavily corrupted by its spurious solution \{-\{\phi^0\}\} since the system is nonsingular (see point 2). On subsequent iterations, the \{D^\prime\} matrix will be treated as if singular, and the modified Nelson's method is used to find the particular solution (see points 3 and 4).

Point 2: Deletion of Spurious Solution
The spurious solution for \{\Delta \phi\} can be seen in Eq. (3), where it is evident that \{\Delta \phi\} = -\{\phi^0\} trivially satisfies the equality. When \{D^\prime\} is only nearly singular due to an approximate \lambda_i, the solution for \{\Delta \phi\} is driven strongly toward this spurious solution. It is only when \{D^\prime\} is exactly singular (convergent \lambda_i) that a meaningful \{\Delta \phi\} is achievable.

The easiest way (and so far the best way) to delete the unwanted spurious solution is to simply add \{\phi^0\} to the corrupted \{\Delta \phi\} solution and scale the result to satisfy perturbed mass normalization (see Ref. 6 for discussion of mass normalization schemes). An alternative method would be to orthogonalize the \{\Delta \phi\} solution to \{\phi^0\} by using a Gram-Schmidt orthogonalization process, but this has not been as effective.

The deletion of the spurious solution has only been necessary on the starting iteration in which \{\Delta \phi\} is found from standard matrix decomposition techniques (nonsingular \{D^\prime\}). It is, however, always an option if convergence difficulties are encountered on subsequent iterations.

Point 3: Homogeneous and Particular Solutions
As for a nonhomogeneous differential equation, the total solution for \{\Delta \phi\} is made up of homogeneous and particular solutions. When \{D^\prime\} is exactly singular, \{\phi^0\} is a homogeneous solution for \{\Delta \phi\} in Eq. (17) since \{D^\prime\} \{\phi^0\} = \{0\}. The corresponding particular solution will be called \{V\}. The total solution for \{\Delta \phi\} is then a sum of \{V\}, and a weighted \{\phi^0\}. The weighting factor \phi_i is introduced because the scaling of the homogeneous solution is initially indeterminate:
For singular \{D^\prime\}:
\[\{\Delta \phi\} = c_i \{\phi^0\} + \{V\},\]
(20)
The following derivation will employ conditions of mass normalization to determine the weighting factor \phi_i. Equation (20) must be altered since \{\phi^0\}, is unknown:
\[\{\Delta \phi\} = c_i \{\phi^0\} + \{V\},\]
(21)
\[(1 - c_i)\{\Delta \phi\} = c_i \{\phi^0\} + \{V\},\]
(22)
\[\{\Delta \phi\} = \frac{c_i}{1 - c_i} \{\phi^0\} + \frac{1}{1 - c_i} \{V\},\]
(23)
A concern at this point is whether or not \phi_i can attain the value of unity, since the solution of Eq. (23) would be undefined. In practice, \phi_i is much smaller than unity, meaning that the contribution of the homogeneous solution to the total solution is small. This fact can be shown with a heuristic proof by contradiction. Letting \phi_i = 1 in Eq. (22) cancels the \{\phi^0\} terms, leaving the particular solution equal to the spurious solution \{V\} = \{-\{\phi^0\}\}. But the spurious solution is deleted in the starting iteration, and the particular solution should not be corrupted with the spurious solution. Therefore, \phi_i must be different than unity.

The mass normalization criteria for \{\phi^0\}, and \{\phi^0\}, are
\[\{\phi^0\}^T \{M^0\} \{\phi^0\} = 1\]
(24)
\[\{\phi^0\}^T \{\phi^0\} = 1\]
(25)

Subtracting the two equations and expanding yields
\[(\{\phi^0\} + \{\Delta \phi\})^T \{M^0\} \{\phi^0\} + \{\Delta \phi\}^T \{\Delta M\} \{\phi^0\} + \{\Delta \phi\}^T \{\phi^0\} = 0\]
(26)
\[2\{\Delta \phi\}^T \{M^0\} \{\phi^0\} + \{\Delta \phi\}^T \{\Delta M\} \{\phi^0\} + 2\{\phi^0\}^T \{\phi^0\} \{\phi^0\} = 0\]
(27)
Combining terms gives the exact normalization equation for \( \Delta \phi \):

\[
2(\Delta \phi)^T \{ M \} (\phi^0) + (\Delta \phi)^T \{ M \} (\Delta \phi) = -(\phi^0)^T \{ M \} (\phi^0) \tag{28}
\]

Substituting Eq. (23) into Eq. (28) for \( \Delta \phi \), gives, after some algebraic manipulation,

\[
\xi_i + (2 + \xi_i) c_i + c_i^2 - c_i^2 = 0 \tag{29}
\]

where

\[
\xi_i = (\phi^0)^T \{ M \} (\phi^0) + 2(\phi^0)^T \{ M \} (V) + (V)^T \{ M \} (V) \tag{30}
\]

Equation (29) has roots

\[
c_i = -1, \quad 1 + \sqrt{1 + \xi_i} \tag{31}
\]

The proper root is the one for which \(|c_i| \ll 1\), i.e.

\[
c_i = 1 - \sqrt{1 + \xi_i} \tag{32}
\]

Point 4: Calculation of Particular Solution

The only remaining problem is the most difficult one, the calculation of the particular solution \( \{ V \} \). This step is not needed on the starting iteration, when \( [D'] \) is nonsingular.

A brief review of Nelson’s method is necessary at this point. The objective is to solve for the particular solution \( \{ V \} \) of Eq. (9) by directly removing the singularity in \( [Z]^0 \). This is accomplished by “zeroing out” a row and column of \( [D]^0 \), putting a 1 at the row and column intersection, and placing a zero in the same row of \( [F]^0 \), effectively enforcing a zero value on the corresponding component of \( \{ V \} \). The row and column chosen is essential to the success of the scheme and corresponds to the maximum magnitude component of the fifth baseline eigenvector. For example, if the fifth component of \( \{ \phi^0 \} \) has maximum magnitude, then one puts zeros on the fifth row and column of \( [D]^0 \), and the fifth row of \( [F]^0 \), and puts a 1 at \( D_{55}^0 \). This drives the fifth component of \( \{ V \} \) to zero in the subsequent matrix decomposition. This method is widely used and is efficient when the coefficient matrix is exactly singular. When the matrix is only nearly singular, however, a modification must be made.

The modified Nelson’s method developed here has two differences from the original method. First, one solves for the particular solution of Eq. (17) instead of Eq. (9). This is a trivial point since Nelson’s method would work equally well on either equation as long as both coefficient matrices were exactly singular. The second change handles the nonsingularity of \( [D] \).

Instead of enforcing a zero value on the component of \( \{ V \} \), we will prescribe a finite value derived from previous \( \{ \Delta \phi \} \) information. On iteration 0, \( \{ \Delta \phi \} \) was found from a standard matrix decomposition. Thus, we retain more of the physics of the problem if we set the component of \( \{ V \} \), equal to the value of the corresponding component of \( \{ \Delta \phi \} \), found on iteration 0. The difference is given in the following equations:

Nelson’s method:

\[
V_{5i} = 0 \tag{33}
\]

Modified Nelson’s method:

\[
V_{5i} = \Delta \phi_{5i} \quad \text{(from iteration 0)} \tag{34}
\]

where the asterisk denotes the component of interest discussed above. Computationally, this makes a monumental difference in the ability of the iterations to converge. Note that the finite change prescribed does not need to be updated since it becomes less important as \( [D] \) becomes more singular; that is, it is just a convergence catalyst and does not influence the final answer.

Solution Procedure

The points addressed earlier each fall neatly into place in the solution algorithm. The algorithm is shown schematically in Fig. 1. The 0th iteration is the starting iteration in which an initial guess on the eigenvalue changes (\( \Delta \lambda \)) is obtained from Fox and Kapoor’s method, and the \( [D]^0 \) matrices are decomposed in the standard way (Nelson’s method is unnecessary) to get the first estimate on the eigenvector changes (\( \{ \Delta \phi \} \)). The eigenvector change on this starting iteration must have the spurious solution removed by adding \( (\phi^0)^T \), and then must be scaled to satisfy the perturbed mass normalization. Subsequent iterations use the full nonlinear update for \( \Delta \lambda \) [Eq. (14)] and the modified Nelson’s method for solving Eq. (17) for its particular solution \( \{ V \} \). Removing the spurious solution has not been necessary on these iterations. Equation (32) is then solved for the weighting factor \( c_i \) and \( \{ \Delta \phi \} \) is updated from Eq. (23). Note that whenever an equation calls for a perturbed value (e.g., \( \lambda^0 \)), the sum of the baseline and the most current perturbation (e.g., \( \lambda^0 + \Delta \lambda \)) is used.

With the essential equations derived and the solution methodology outlined, the focus will now turn to solving an example problem.

Benchmark Problem: Cantilever Beam

Test Problem

A five-element Euler-Bernoulli cantilever beam model (Fig. 2) is proposed as a benchmark problem due to its simplicity in finite element analysis. The important material constants are \( E = 2.0684 \times 10^7 \) MPa, \( \nu = 0.3 \), \( \rho = 7.8334 \times 10^{-9} \) N s^2/m^4. The motion is constrained to allow only \( \theta \) bending, which leaves a total of 10 degrees of freedom (\( z \) displacement and \( \theta \) rotation at each of five nodes). Element stiffness and mass matrices are assembled and used to find the baseline eigenpairs through the use of a simple MATLAB program (MATLAB is a high-level matrix manipulation language that pro-
vides easy access to the matrix software developed by the LINPACK and EISPACK projects. Upon prescription of design variable changes, one computes new stiffness and mass matrices and Fox and Kapoor’s method is performed to get an initial estimate of the eigenvalue perturbations. If the prescribed changes are large, this initial estimate will be in error and the nonlinear iteration scheme described herein must be performed. The aforementioned 100% reduction in an element’s stiffness cannot be tested here since the structure is not redundant.

Several measures have been developed to monitor the progress of the iterations. First, convergence of the eigenvalues is shown through the average of the absolute percent error in each \( \lambda_i \), where

\[
\text{absolute percent error} = \left| \frac{\lambda_i - \lambda_{i, \text{exact}}}{\lambda_{i, \text{exact}}} \right| \times 100
\]

and \( i \) ranges from 1 to 10 (Table 2 heading: average [\% error] in \( \lambda_i \)). Second, the increasing singularity of \( [D^1] \) is shown through the average condition numbers of these matrices (Table 2 heading: average \([D^1]\) cond no.). Finally, convergence of the \( [\phi^1] \) vectors is monitored through an average of the Euclidean norms of \( [\phi^1] - [\phi^1]_{\text{exact}} \) (Table heading: average \([\phi^1]\) error II).

Five design variable scaling sets are to be analyzed, each of which perturbs all design variables simultaneously. This is not the typical process for calculating sensitivities, where only one design variable is perturbed. Simultaneous perturbations were performed to test the algorithm, since obtaining convergent results will obviously be more difficult than for a single design variable perturbation. The five design variable scalings are shown in Table 1 and the design variables (abbreviated DV) being scaled are the corresponding beam element heights.

The first set corresponds to case 1, subcase 3 of the cantilever beam work of Kim et al. It was chosen to obtain verification results. The other four design variable scaling sets were chosen arbitrarily to have varying signs and similar magnitude scalings. The last column shows the number of converged eigenpairs.

The results and discussion will be discussed in the following section.

Results and Discussion

Table 2 shows the iteration history for design variable scaling set 1. Improvement of more than an order of magnitude can be seen in all three measures of convergence for each iteration following the first.

To give meaning to these convergence measures, detailed data will be supplied for iteration 3 shown in Table 2. In Table 3, the perturbed eigenvalues at iteration 3 are compared with the baseline eigenvalues, the perturbed eigenvalues returned by Fox and Kapoor’s modal method (iteration 0), and the exact perturbed eigenvalues.

Table 1 Design variable scaling sets

<table>
<thead>
<tr>
<th>Set</th>
<th>Magnitude, %</th>
<th>DV1, %</th>
<th>DV2, %</th>
<th>DV3, %</th>
<th>DV4, %</th>
<th>DV5, %</th>
<th>Converged eigenpairs</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>\pm 20</td>
<td>+ 19.15</td>
<td>- 2.89</td>
<td>- 9.01</td>
<td>- 6.34</td>
<td>- 8.27</td>
<td>10</td>
</tr>
<tr>
<td>2</td>
<td>\pm 25</td>
<td>+ 19.15</td>
<td>- 17.89</td>
<td>+ 19.01</td>
<td>- 18.34</td>
<td>- 17.27</td>
<td>10</td>
</tr>
<tr>
<td>3</td>
<td>\pm 30</td>
<td>+ 25</td>
<td>- 24</td>
<td>+ 24.5</td>
<td>- 23</td>
<td>+ 24.7</td>
<td>10</td>
</tr>
<tr>
<td>4</td>
<td>\pm 35</td>
<td>+ 28.5</td>
<td>- 29</td>
<td>+ 27</td>
<td>- 28</td>
<td>+ 30</td>
<td>10</td>
</tr>
<tr>
<td>5</td>
<td>\pm 40</td>
<td>+ 39.15</td>
<td>- 37.89</td>
<td>+ 39.01</td>
<td>- 38.34</td>
<td>+ 37.27</td>
<td>9</td>
</tr>
</tbody>
</table>

Table 2 Convergence of iteration — moderate change

<table>
<thead>
<tr>
<th>Iter. no.</th>
<th>Average [% error] in ( \lambda_i )</th>
<th>Average ([D^1]) cond no.</th>
<th>Average ([\phi^1]) error II</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>3.8161 Fox</td>
<td>4.2551e-06</td>
<td>1.9177</td>
</tr>
<tr>
<td>1</td>
<td>3.9633</td>
<td>4.0960e-06</td>
<td>0.8927</td>
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<td>0.1386</td>
<td>1.4377e-07</td>
<td>0.0352</td>
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<td>3</td>
<td>0.0055</td>
<td>7.6016e-09</td>
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<td>4.2493e-10</td>
<td>7.3449e-5</td>
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<td>1.1735e-5</td>
<td>2.4589e-11</td>
<td>3.9077e-6</td>
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<td>6</td>
<td>5.9988e-7</td>
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<td>2.1835e-7</td>
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<td>7</td>
<td>3.1942e-8</td>
<td>8.5910e-14</td>
<td>1.2547e-8</td>
</tr>
</tbody>
</table>

Table 3 Convergence of eigenvalues — moderate change

<table>
<thead>
<tr>
<th>Mode no.</th>
<th>( \lambda_i )</th>
<th>Fox and Kapoor’s ( \lambda_i ) (iter. 0)</th>
<th>Exact ( \lambda_i )</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>6.8007e + 04</td>
<td>9.7977e + 04</td>
<td>9.0064e + 04</td>
</tr>
<tr>
<td>2</td>
<td>2.6735e + 06</td>
<td>2.9831e + 06</td>
<td>2.8705e + 06</td>
</tr>
<tr>
<td>3</td>
<td>2.1091e + 07</td>
<td>2.1859e + 07</td>
<td>2.1091e + 07</td>
</tr>
<tr>
<td>4</td>
<td>8.2307e + 07</td>
<td>8.3195e + 07</td>
<td>8.3556e + 07</td>
</tr>
<tr>
<td>5</td>
<td>2.2674e + 08</td>
<td>2.2991e + 08</td>
<td>2.1670e + 08</td>
</tr>
<tr>
<td>6</td>
<td>6.2576e + 08</td>
<td>6.2836e + 08</td>
<td>6.0003e + 08</td>
</tr>
<tr>
<td>7</td>
<td>1.3384e + 09</td>
<td>1.3149e + 09</td>
<td>1.2824e + 09</td>
</tr>
<tr>
<td>8</td>
<td>2.8149e + 09</td>
<td>2.6987e + 09</td>
<td>2.6743e + 09</td>
</tr>
<tr>
<td>9</td>
<td>5.6807e + 09</td>
<td>5.1008e + 09</td>
<td>5.0917e + 09</td>
</tr>
<tr>
<td>10</td>
<td>1.2293e + 10</td>
<td>1.0553e + 10</td>
<td>1.0406e + 10</td>
</tr>
</tbody>
</table>

Table 4 Convergence of eigenvectors — moderate change

\( [\phi^1]_16 \) Fox and Kapoor’s \( [\phi^1]_16 \) (iter. 3) Exact \( [\phi^1]_16 \)

<table>
<thead>
<tr>
<th>Iter. no.</th>
<th>Average [% error] in ( [\phi^1] )</th>
<th>Average ([D^1]) cond no.</th>
<th>Average ([\phi^1]) error II</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>-3.9201</td>
<td>-4.8353</td>
<td>-4.4618</td>
</tr>
<tr>
<td>1</td>
<td>-0.1392</td>
<td>-0.0804</td>
<td>-0.0863</td>
</tr>
<tr>
<td>2</td>
<td>1.5953</td>
<td>3.7807</td>
<td>3.4092</td>
</tr>
<tr>
<td>3</td>
<td>0.1631</td>
<td>0.1603</td>
<td>0.1530</td>
</tr>
<tr>
<td>4</td>
<td>0.7704</td>
<td>-0.8367</td>
<td>-0.7196</td>
</tr>
<tr>
<td>5</td>
<td>-0.1675</td>
<td>-0.1885</td>
<td>-0.1868</td>
</tr>
<tr>
<td>6</td>
<td>-2.8386</td>
<td>-2.3515</td>
<td>-2.4258</td>
</tr>
<tr>
<td>7</td>
<td>0.1452</td>
<td>0.1715</td>
<td>0.1711</td>
</tr>
<tr>
<td>8</td>
<td>10.459</td>
<td>11.347</td>
<td>11.443</td>
</tr>
<tr>
<td>9</td>
<td>-0.2072</td>
<td>-0.2296</td>
<td>-0.2312</td>
</tr>
<tr>
<td>10</td>
<td>-0.2072</td>
<td>-0.2296</td>
<td>-0.2312</td>
</tr>
</tbody>
</table>

Table 5 Convergence of iteration — large change

<table>
<thead>
<tr>
<th>Iter. no.</th>
<th>Average [% error] in ( \lambda_i )</th>
<th>Average ([D^1]) cond no.</th>
<th>Average ([\phi^1]) error II</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>76.3080 Fox</td>
<td>2.0894e-05</td>
<td>4.1988</td>
</tr>
<tr>
<td>1</td>
<td>73.5623</td>
<td>1.7626e-05</td>
<td>11.0364</td>
</tr>
<tr>
<td>2</td>
<td>19.7245</td>
<td>5.9050e-06</td>
<td>15.1258</td>
</tr>
<tr>
<td>3</td>
<td>5.3836</td>
<td>3.1959e-06</td>
<td>3.5353</td>
</tr>
<tr>
<td>4</td>
<td>4.9042</td>
<td>2.4621e-06</td>
<td>1.9003</td>
</tr>
<tr>
<td>5</td>
<td>3.5970</td>
<td>1.5109e-06</td>
<td>2.9056</td>
</tr>
<tr>
<td>6</td>
<td>4.8248</td>
<td>1.6387e-06</td>
<td>1.4242</td>
</tr>
<tr>
<td>7</td>
<td>3.4145</td>
<td>1.0459e-06</td>
<td>2.6031</td>
</tr>
</tbody>
</table>

Averaging the percent errors for Fox and Kapoor’s \( \lambda_i \) (iter. 0) and \( \lambda_i \) (iter. 3) in Table 5 gives rise to the 3.8% and 0.0055% figures in Table 2.
the most poorly behaved eigenvector for this design variable scaling set (0.0091 vector norm error in iteration 3 estimate; average vector norm error for the 10 vectors was 0.0015). Nodal displacements and rotations alternate in the eigenvector component sequence.

It is evident that the perturbed mode shape estimate at iteration 3 is much improved over the mode shape from Fox and Kapoor’s modal method, even for the most slowly converging eigenvector.

For larger design variable perturbations, convergent results (not tabulated) were obtained, but not as quickly. For design variable scaling set 2, six iterations were necessary to converge from a 13% average error in λ₁ (Fox and Kapoor’s modal method), a 1.0e-05 average condition number in [D'], and a 3.4 average column norm of [A₀] error to the accuracy of iteration 3 earlier. For the larger perturbations of design variable scaling set 3, eight iterations were necessary to converge from a 24% average error in λ₁ (Fox and Kapoor’s method), a 1.4e-05 average condition number, and a 2.6 average column norm of [A₀] error to the accuracy of iteration 3 earlier. For design variable scaling set 4, oscillations in the iterations began to appear, but the convergence criteria still improved from 35% (Fox and Kapoor) to 24% average error in X- (Fox and Kapoor’s modal method), a 1.4e-05 average condition number, and a 2.6 average column norm of [D₀] error to the accuracy of iteration 3 earlier. For design variable scaling set 5, oscillations in the iterations began to appear, but the convergence criteria still improved from 35% (Fox and Kapoor) to 24% average error in X- (Fox and Kapoor’s modal method), a 1.4e-05 average condition number, and a 2.6 average column norm of [D₀] error to the accuracy of iteration 3 earlier.

Convergent criteria for all 10 eigenpairs have been shown for a 310% increase in stiffness for the perturbed value of the eigenvalue and eigenvector, and the finite change in the eigenvector is dependent on the perturbed value of the eigenvalue and eigenvector, respectively. This method is particularly valuable in the case when destruction of a component of a large structure is a possibility.

Table 6 Severe design variable scaling sets

<table>
<thead>
<tr>
<th>Set</th>
<th>Magnitude, %</th>
<th>DV1, %</th>
<th>DV2, %</th>
<th>DV3, %</th>
<th>DV4, %</th>
<th>DV5, %</th>
<th>Nonconv. eigensets</th>
</tr>
</thead>
<tbody>
<tr>
<td>6</td>
<td>±45</td>
<td>+45</td>
<td>-44</td>
<td>+43</td>
<td>-44.4</td>
<td>+44.7</td>
<td>5th</td>
</tr>
<tr>
<td>7</td>
<td>±50</td>
<td>+49</td>
<td>-48.5</td>
<td>+49.4</td>
<td>-48.4</td>
<td>+50</td>
<td>5th</td>
</tr>
<tr>
<td>8</td>
<td>±45</td>
<td>+45</td>
<td>+44</td>
<td>+43</td>
<td>-44.4</td>
<td>-44.7</td>
<td>9th, 10th</td>
</tr>
<tr>
<td>9</td>
<td>±50</td>
<td>+49</td>
<td>+48.5</td>
<td>+49.4</td>
<td>-48.4</td>
<td>-50</td>
<td>9th, 10th</td>
</tr>
<tr>
<td>10</td>
<td>none</td>
<td>0</td>
<td>0</td>
<td>+60</td>
<td>0</td>
<td>1</td>
<td>none</td>
</tr>
<tr>
<td>11</td>
<td>none</td>
<td>0</td>
<td>0</td>
<td>+80</td>
<td>0</td>
<td>0</td>
<td>10th</td>
</tr>
<tr>
<td>12</td>
<td>none</td>
<td>0</td>
<td>0</td>
<td>+80</td>
<td>0</td>
<td>0</td>
<td>10th</td>
</tr>
</tbody>
</table>

Does not apply.

Conclusions

A robust solution of the full nonlinear eigenproblem perturbation equations has been developed. No terms in the equations for Δλ and {Δφ} have been neglected, and the frequency and mode shape perturbation equations are coupled (unlike the first-order system). That is, the finite change in eigenvalue is dependent on the perturbed value of the eigenvector, and the finite change in the eigenvector is dependent on the perturbed value of the eigenvalue. An iterative solution is therefore required. The familiar singularity in the solution for the eigenvector change is treated using the authors’ modification of Nelson’s method for resolution of the singular matrix equations. The normalization task necessary for {Δφ} also retains all terms, leading to a cubic equation for a weighting factor c₀.

In a cantilever beam example, convergence to the exact answers for eigenvalue and eigenvector changes has been shown to be achievable for very large perturbations in design variables. Fast convergence has been shown for 100% increases in stiffness and near convergence has been shown for 170% increases in stiffness, for the case of simultaneous design variable perturbations. For the case of a single design variable perturbation, convergence of all 10 eigenvectors has been shown for a 310% increase in stiffness for the perturbed beam element. This makes the proposed scheme especially valuable in the case when destruction of a component of a large structure is a possibility.

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References


