

Method of unsteady aerodynamic forces approximation for aeroservoelastic interactions

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Introduction

The adverse interactions occurring between the three main disciplines : unsteady aerodynamics, aeroelasticity and servo-controls are called aeroservoelastic interactions. These interactions can be described mathematically by a system of equations in a state space form. This system requires a different representation for the unsteady aerodynamic forces from that for the classical flutter equation. The unsteady aerodynamic forces, in the case of the classical flutter equation or aeroelasticity, are calculated by the Doublet Lattice Method (DLM) in the frequency domain for a set of reduced frequencies k 's and Mach numbers M 's. The error in the DLM modelling due to omitting the effect of the boundary layer, viscosity, flow separation, etc are of higher order than the error in curve fitting frequency domain modal generalized forces by $j\omega$ rational functions and replacing $j\omega$ by s . Since time domain Linear Time Invariant Ordinary Differential Equations (LTI ODE) are required for using modern control theory design, several approximations for the unsteady aerodynamic forces

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in the s -domain have been developed. There are mainly three formulations to approximate the unsteady generalized forces by rational functions in the Laplace domain in the frequency domain¹⁻⁴ : Least Square (LS), Modified Matrix Padé (MMP) and Minimum State (MS). The approximation yielding the smallest order time domain LTI state-space model is the MS⁵ approximation method. The dimension of the time domain LTI state-space model depends on the number of retained modes and the number of aerodynamic lags n_a . There is a trade-off between the number of aerodynamic lags and the accuracy of the approximation. The higher the n_a , the better the approximation, but the order of the time domain LTI state-space model is larger. The order of the LTI state-space model strongly affects the efficiency of subsequent analyzes. In the present paper, a new method for the determination of efficient state-space aeroservoelastic models is presented. This method combines results from the theory of Padé approximants and the theory of model reduction developed in the frame of control theory. Finally, a comparison between the new method and the Minimum State MS method is presented. The error of our method is 12 to 40 times smaller than the error of the MS approximation method for the same number of augmented states n_a , and depends on the choice of the model reduction method.

Aircraft equations of motion

The motion of an aircraft modelled as a flexible structure with no forcing terms is described by the following equations, written in the time-domain:

$$\tilde{\mathbf{M}} \ddot{\eta} + \tilde{\mathbf{C}} \dot{\eta} + \tilde{\mathbf{K}} \eta + q_{dyn} \mathbf{Q}(k) \eta = 0. \quad (1)$$

Here η is the generalized variable defined as $\mathbf{q} = \Phi \eta$, where q is the displacement vector and Φ is the matrix formed with the eigenvectors of the free-vibration problem $\mathbf{M} \ddot{\mathbf{q}} + \mathbf{K} \mathbf{q} = 0$. Moreover $\tilde{\mathbf{M}} = \Phi^T \mathbf{M} \Phi$, $\tilde{\mathbf{C}} = \Phi^T \mathbf{C} \Phi$, $\tilde{\mathbf{K}} = \Phi^T \mathbf{K} \Phi$ and $\mathbf{Q}(k) = \Phi^T \mathbf{A}_e(k) \Phi$, where \mathbf{M} , \mathbf{K} and \mathbf{C} are the generalized mass, the elastic stiffness and the damping matrices, $q_{dyn} = 0.5 \rho V^2$ is the dynamic pressure, where ρ is the air density, V is the true airspeed, $k = \frac{\omega b}{V}$ is the reduced frequency, ω is the natural frequency, b is the wing semichord length, $\mathbf{A}_e(k)$ is the aerodynamic influence coefficient matrix for a given Mach number M and a set of $k \in \{k_1, k_2, \dots, k_p\}$ values.

Applying the Laplace transform to equation (1) we obtain

$$[\tilde{\mathbf{M}} s^2 + \tilde{\mathbf{C}} s + \tilde{\mathbf{K}}] \eta(s) + q_{dyn} \mathbf{Q}(s) \eta(s) = 0 \quad (2)$$

The approximation of the unsteady aerodynamic forces is a necessary prerequisite to the control analysis of the subsequent aeroelastic system. Since $\mathbf{Q}(k)$ can only be tabulated for a finite set of reduced frequencies, at a fixed Mach number M , it must be interpolated in the s -domain in order to obtain $\mathbf{Q}(s)$. Assuming analyticity of \mathbf{Q} over the region of interest one could find the Laurent series for $\mathbf{Q}(s) = \sum c_i s^i$. This approach is not desirable from a control point of view since it introduces derivatives of order greater than two. Since we expect the linear aeroelastic system to behave as a Bounded Input Bounded Output (BIBO) system it would seem obvious to find a Rational Fraction Approximation (RFA) for $\mathbf{Q}(s)$. The method giving the lowest aerodynamic dimension is the MS approximation⁵ :

$$\mathbf{Q}^{MS}(s) = \mathbf{A}_0 + \mathbf{A}_1 s + \mathbf{A}_2 s^2 + \mathbf{D} (s \mathbf{I} - \mathbf{R})^{-1} \mathbf{E}, \quad (3)$$

In the last term of the above equation, the denominator coefficients are the same as the ones considered in the LS method and the numerator are calculated as a coupled product of \mathbf{D} and \mathbf{E} matrices. The diagonal matrix of aerodynamic roots is $\mathbf{R} = \text{diag}(b_1, \dots, b_p)$, where the b_i 's are the lags which are usually chosen to belong in the range between zero and the highest frequency of vibration.

For a given set of lags b_p , the MS numerator coefficient matrices \mathbf{D} and \mathbf{E} , are determined using an iterative, nonlinear least square LS methods that minimize an overall error function J_{MS} , see¹.

The iterative technique assumes an initial \mathbf{D} , calculates \mathbf{E} in order to calculate a new \mathbf{D} matrix using LS methods. This iterative technique⁵ continues until coefficient matrices are found with a converged minimum error.

Rational Function Approximation Method

In this section we present a new method for approximating the unsteady aerodynamic forces. Let us denote by \mathbf{Q}_i^{TAB} the matrix formed with the tabulated value of the element $\mathbf{Q}_{ij}(\mathbf{i} k_l)$, where \mathbf{i} denotes the pure imaginary number and k_l belongs to the set of reduced frequencies for which the DLM method is available. We approximate each element of the unsteady aerodynamic influence matrix by a $[N_{ij} + 2, N_{ij}]$ Padé approximation, where N_{ij} is a natural number depending on the element to be approximated and chosen in such way that we have a good accuracy for the approximation. Denoting by s the Laplace variable, we have

$$\bar{\mathbf{Q}}_{ij}(s) = \frac{P_{N+2}(s)}{R_N(s)} = \frac{a_0 s^{N+2} + a_1 s^{N+1} + \dots + a_{N+2}}{b_0 s^N + b_1 s^{N-1} + \dots + b_N}, \quad (4)$$

where the indices ij for the coefficients of the polynomials P_{N+2} and R_N have been omitted. For details on Padé approximants see⁶.

Since the aeroelastic system should be a BIBO system we will impose that the coefficients of the denominator satisfy the Routh-Hurwitz criterion. The coefficients of the Padé approximants are found by using a standard Least-Square LS technique, and by minimizing the following error function:

$$J_{ij} = \left(\sum_{p=1}^l | [\mathbf{Q}_p^{TAB}]_{ij} - \bar{\mathbf{Q}}_{ij}(\mathbf{i}k_p) |^2 \right)^{1/2}, \quad (5)$$

for each couple (i, j) . Here $|\cdot|$ denotes the modulus of a complex number. Once the coefficients of P_{N+2} and R_N are determined, it is a simple matter to write the aerodynamic approximation $\bar{\mathbf{Q}}$ as follows:

$$\bar{\mathbf{Q}}(s) = \mathbf{A}_0 + \mathbf{A}_1 s + \mathbf{A}_2 s^2 + \mathbf{Z}(s) s, \quad (6)$$

where $\mathbf{Z}(s)$ is a proper rational matrix. Indeed, each element of $\bar{\mathbf{Q}}_{ij}$ is the ratio of a polynomial of degree $N + 2$ and a polynomial of degree N . For an arbitrary fixed couple of indices (i, j) , let us find $\alpha_{0,ij}, \alpha_{1,ij}, \alpha_{2,ij}$ and $c_{0,ij}, c_{1,ij}, \dots, c_{N-1,ij}$ such that, for all s we will write equation (6) (for each couple (i, j)), as follows:

$$Q_{ij}(s) = \frac{a_0 s^{N+2} + \dots + a_{N+2}}{b_0 s^N + \dots + b_N} = \alpha_0 s^2 + \alpha_1 s + \alpha_2 + \frac{c_0 s^{N-1} + \dots + c_{N-1}}{b_0 s^N + b_1 s^{N-1} + \dots + b_N} s.$$

We have dropped the indices ij in order to simplify the notation. By identification we obtain the following linear system:

$$\begin{pmatrix}
1 & 0 & 0 & 0 & 0 & \dots & 0 \\
b_1 & 1 & 0 & 0 & 0 & \dots & 0 \\
b_2 & b_1 & 1 & 1 & 0 & \dots & 0 \\
b_3 & b_2 & b_1 & 0 & 1 & \dots & 0 \\
b_4 & b_3 & b_2 & 0 & 0 & \dots & 0 \\
\dots & \dots & \dots & \dots & \dots & \dots & \dots \\
0 & b_N & b_{N-1} & 0 & 0 & \dots & 1 \\
0 & 0 & b_N & 0 & 0 & \dots & 0
\end{pmatrix}
\begin{pmatrix}
\alpha_0 \\
\alpha_1 \\
\alpha_2 \\
c_0 \\
c_1 \\
\dots \\
c_{N-2} \\
c_{N-1}
\end{pmatrix}
=
\begin{pmatrix}
a_0 \\
a_1 \\
a_2 \\
a_3 \\
a_4 \\
\dots \\
a_{N+1} \\
a_{N+2}
\end{pmatrix}$$

From the first, the second and the last equation we obtain $(\alpha_i)_{i=\overline{0,2}}$. By back-substitution, the values of $(c_i)_{i=\overline{0,N-1}}$ are found.

It is easy to see that the determinant of the previous system is b_N and b_N is different from zero since it equals the product of the roots of $R_N(s)$ which have strictly negative real parts.

For every couple (i, j) , we define $[\mathbf{A}_p]_{ij} = \alpha_{2-p,ij}$ with $p = \overline{0,2}$ and

$$\mathbf{Z}_{ij}(s) = \frac{c_{0,ij}s^{N-1} + c_{1,ij}s^{N-2} + \dots + c_{N-1,ij}}{b_{0,ij}s^N + b_{1,ij}s^{N-1} + \dots + b_{N,ij}} \quad (7)$$

With these new definitions we find $\bar{\mathbf{Q}}(s)$ by equation (6).

$\mathbf{Z}(s)$ is a strictly proper rational matrix and therefore can be viewed as the transfer function of a linear system. We wish to find a triple $(\mathbf{A}, \mathbf{B}, \mathbf{C})$ such that we have $\mathbf{Z}(s) \approx \mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{B}$. This can be achieved either by constructing a minimal order realization (denoted by Minreal) or by constructing a reduced model starting from a known realization

(for example the canonical or the modal realization). The first approach gives an equality sign, while the second approach can be viewed as an approximation in a chosen norm.

A model order reduction can be described as follows: Starting with a full order model $\mathbf{Z}(s)$, find a lower order model $\hat{\mathbf{Z}}(s)$ such that \mathbf{Z} and $\hat{\mathbf{Z}}$ are close (where the closeness is to be defined). This can be achieved either by an *additive* model order reduction or by a *relative-multiplicative* model order reduction. The additive model order reduction consists in finding $\hat{\mathbf{Z}}$ such that, $\mathbf{Z} = \hat{\mathbf{Z}} + \Delta_{add}$, where Δ_{add} is small in some norm.

The relative-multiplicative model order reduction comes to finding $\hat{\mathbf{Z}}$ such that: $\hat{\mathbf{Z}} = \mathbf{Z}(I + \Delta_{rel})$ and $\mathbf{Z} = \hat{\mathbf{Z}}(I + \Delta_{rel})$ where I is the identity and Δ_{rel} is small in some norm. The two additive methods available in the Matlab Robust Toolbox used in this article are:

- Schur balanced model order reduction (Schur).
- Optimal Hankel approximation (Ohkapp).

The relative-multiplicative method is

- Balanced stochastic truncation (Bst-Rem) with relative error model reduction.

For details concerning this methods, see the MATLAB documentation and the references therein.

We can therefore proceed as follows:

1. By eliminating all the unobservable and uncontrollable states of the linear system whose transfer function is given by $\mathbf{Z}(s)$, we construct a minimal-order transfer equivalent realization which gives a perfect match, i.e. $\mathbf{Z}(s) = \mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{B}$.

2. Finding $\hat{\mathbf{Z}}(s) = \hat{\mathbf{C}} (s\mathbf{I} - \hat{\mathbf{A}})^{-1} \hat{\mathbf{B}}$ such that $\|\hat{\mathbf{Z}}(s) - \mathbf{Z}(s)\|_\infty \leq \varepsilon$, where $\varepsilon > 0$ is a small tolerance and $\|\cdot\|_\infty$ is the \mathbf{H}^∞ norm, we construct an optimal approximation of $\mathbf{Z}(s)$.

If we denote by $\hat{\mathbf{Q}}(s) = \mathbf{A}_0 + \mathbf{A}_1 s + \mathbf{A}_2 s^2 + \hat{\mathbf{Z}}(s) s$, then we have

$$\|\hat{\mathbf{Q}}(s) - \mathbf{Q}^{TAB}\|_\infty \leq \|\hat{\mathbf{Q}}(s) - \bar{\mathbf{Q}}(s)\|_\infty + \|\bar{\mathbf{Q}}(s) - \mathbf{Q}^{TAB}\|_\infty \leq \|\bar{\mathbf{Q}}(s) - \mathbf{Q}^{TAB}\|_\infty + \varepsilon.$$

This shows that the additive model order reduction $\mathbf{Z}(s)$ by $\hat{\mathbf{Z}}(s)$ only degrades the norm of the approximation by an order of ε . A similar inequality can be showed for the relative-multiplicative model order reduction.

Numerical Results

A flexible aircraft with 5, 10, 15 and 20 vibration modes has been considered. The finite element model of the symmetric one half of an aircraft is used to verify this new optimization theory of unsteady aerodynamic forces⁷. The Doublet Lattice Method DLM implemented in STARS⁷ was used to obtain the tabulated unsteady aerodynamic matrices in the frequency domain, for a given Mach number $M = 0.8$ and a set of 14 reduced frequencies $k \in \{0.01, 0.1, 0.2, 0.303, 0.4, 0.5, 0.5882, 0.6250, 0.6667, 0.7143, 0.7692, 0.8333, 0.9091, 1.0000\}$.

With this tabulated data each element of the aerodynamic matrix is approximated by a Padé approximant. The global error of the approximation of our method is reported in the last column of Table 1. Following the procedure developed in the previous section, we construct a strictly proper rational matrix from the matrix of Padé approximants (equation (7)).

Considering the strictly proper rational matrix as the transfer function of a linear system, we construct a reduced order model using different reduction methods from Control Theory.

Eliminating the unobservable and the uncontrollable states with the help of the Minreal function of Matlab we obtain a minimal realization of order n_a , as shown in the second column in Table 1. Following the results of the previous section, we construct the two additive reduced order models, using the Schur and the Optimal Hankel approximation method. The reduced order model for the latter two methods are reported in columns 4 and 6 of Table 1, respectively. The 8th column represents the dimension of the reduced system obtained using the Balanced Stochastic truncation method. We can see that this last method gives the best results in terms of aerodynamic dimension n_a in comparison to the other methods. The tolerance used in the calculation of the minimal realization and the approximation of the reduced order models was chosen to be 10^{-6} . The aerodynamic dimensions n_a found by the four methods are now used to perform the MS approximation of the unsteady aerodynamic forces. This is done in order to compare the errors of unsteady aerodynamic forces approximation for a fixed aerodynamic dimension n_a . The errors for the MS method, denoted by J_{MS} for each method, are reported in columns 3, 5, 7 and 9 of Table 1. It can be seen that the approximation error calculated by our method is 12 to 40 times smaller than the errors calculated by the MS procedure.

Conclusion

The main contribution of this paper is an original method for the approximation of the unsteady aerodynamic forces using recent results from linear system theory and Padé approximants. Starting with a Padé approximation of the unsteady aerodynamic forces we construct a reduced order model for stability analysis purposes. Contrary to the standard MS approximation, the error of the approximation is independent on the aerodynamic dimension

of the final aeroelastic system. Running the MS procedure with different number of lags to get a good approximation can take large amounts of time since the procedure is highly iterative. Our method overcomes the problem of choosing the number of lags (aerodynamic dimension) n_a of the MS procedure, since in our method the aerodynamic dimension is a result not an initial parameter. Furthermore, our method yields better approximation error.

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List of Table Captions

Table 1: J and n_a calculations by different reduction methods

Modes	Minreal		Schur		Ohkapp		Bst-Rem		Error
	n_a	J_{MS}	n_a	J_{MS}	n_a	J_{MS}	n_a	J_{MS}	J
5	17	3.69	19	3.56	18	3.99	6	11.16	0.26
10	81	6.16	80	6.16	79	7.36	31	11.28	0.51
15	192	10.40	187	10.46	82	16.46	37	23.27	0.73
20	344	23.45	337	23.78	152	24.16	69	38.19	1.07