

Aeroservoelastic Optimization of a Transport Aircraft Wingbox under Stochastic Gust Constraints

BRET STANFORD¹

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Abstract

This work considers the aeroservoelastic optimization of a highly flexible transport aircraft wingbox with several control surfaces distributed along the trailing edge. The steady deflections of the control surfaces are designed to alleviate static maneuver loads, while the unsteady deflections are designed to alleviate stochastic continuous gust disturbances. Spatially-detailed unsteady stochastic stress and panel buckling constraints are formulated via modal acceleration, and by methods based on the proximity of the structural design to the equal-probability failure hypersurface spanned by the stochastic gust loads. For the case considered here, it is found that the inclusion of such gust constraints during optimization presents a sizable structural mass increase. In some cases, this mass can be completely recovered with controlled gust load alleviation via linear quadratic controllers driven by multiple flaps distributed along the trailing edge.

Nomenclature

A	dynamic aeroelastic coefficients
A_s, B_s, F_s	static aeroelastic coefficients
B_θ, B_δ, B_w	dynamic aeroelastic input matrices
B_{σ_e}, B_{N_e}	strain displacement matrices
C_p	aerodynamic pressures
C_{σ_e}	stress covariance matrix
F, f	quadratic failure surface terms
$\bar{g}_e$	gust constraint
H	mode acceleration matrix
J	LQ cost function
K_δ	FCS proportional control
K_θ	LQ feedback matrix
KS	aggregation parameter
L	Kalman gain
M, C, K, A	mass, damping, stiffness, and aerodynamic matrices
N_L	number of load cases
N_σ, N_μ	number of stress and buckling aggregation parameters
N_x, N_{xy}	in-plane tractions
q	design variables
Q_x, R_θ	LQ weighting matrices
s_e, s_{e_1}, s_{e_g}	total stress, cruise stress, gust stress

¹ Aeroelasticity Branch, NASA Langley Research Center
Hampton, Virginia, USA, bret.k.stanford@nasa.org

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t, t_s, h_s	patch thickness, stiffener thickness, stiffener height
$\mathbf{u}$	structural solution vector
U_{EAS}, U^*, U_σ	equivalent air speed, critical flutter speed, design gust speed
w, w_g	white noise input, gust speed
$\mathbf{x}$	dynamic aeroelastic solution vector
$\mathbf{x}_s$	static aeroelastic solution vector
$\mathbf{X}$	aeroservoelastic covariance matrix
$\mathbf{y}, \mathbf{v}$	sensor output, sensor noise
α, δ	static trim angles
Θ, θ	commanded and actual control surface rotation vectors
$\sigma_x, \sigma_y, \sigma_{xy}$	plane stresses
σ_Y	yield stress
γ, ω	dynamic aeroelastic eigenvalue, real and imaginary part
ϕ	mode shapes
η	modal amplitudes

1. Introduction

Modern aeroservoelastic optimization typically considers the simultaneous design of multiple subsystems, including the wing jig shape details, the structural wingbox sizing, control surface scheduling, and control law architecture: see for example Refs. [1] - [5], and a broader summary in Ref. [6]. In these papers, articulated control surfaces are used for active cruise drag reduction, maneuver load alleviation (MLA), gust load alleviation (GLA), and active flutter suppression (AFS), among other vehicle-level flight control system (FCS) uses. The way in which this control system should be optimized is tightly coupled to the design decisions made for the structural sizing, jig shape details, etc.

One focus of this work is on the design interplay between the different load cases, particularly between maneuver loads and gust loads, handled either passively or actively. In this work, MLA indicates a situation where the control surfaces optimally distribute the load over the wing while maintaining trim conditions, whereas GLA is a true alleviation in the sense that gust perturbations to a trimmed state are minimized via oscillatory control surface motions.

It should be expected that the inclusion of MLA in the design process will allow an optimizer to feasibly reduce structural weight, since the loads borne by the wingbox will be reduced. This in turn will amplify the role of flutter and gust constraints during the optimization (i.e., make the constraints harder to satisfy), which can in turn be attenuated through the use of AFS and/or GLA. A second focus of this work is the implementation of stochastic (continuous) stress- and buckling-based gust constraints (where linear buckling is considered at the stiffened panel level) for spatially-detailed structural wingbox design, as well as control law design, for a generic transport configuration.

It is well known [6, 7] that continuous gust constraints are efficiently developed using state-space methods driven by a white noise input sent through a forming filter. A Lyapunov equation is then solved for the covariance matrix of each aeroservoelastic state. If the design metrics of interest are linear functions of the state vector (such as structural deflections, accelerations, bending moments, etc.), then the covariances and root-mean-squared values of these metrics are rapidly computed for constraint formulation, as done in Refs. [8], [9], and [10], to name a few. The analytical sensitivities of these terms, typically required for gradient-based optimization, may also be computed [7, 11, 5].

For a large-scale structurally-detailed wingbox with complex failure mechanisms, however, accelerations and bending moments may be too simplistic to

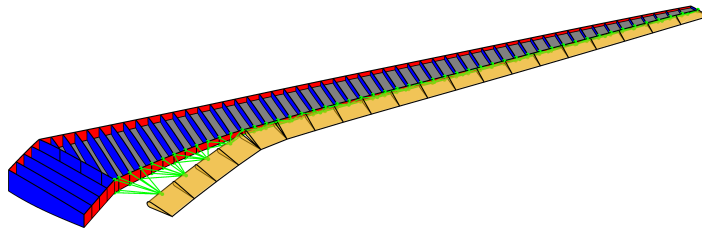


Figure 1: Wingbox and control surfaces of the high aspect ratio CRM.

serve as design constraints. Detailed stress constraints, typically the superposition of $1g$ cruise stresses and the stochastic gust stresses, introduce two numerical challenges. First, the aeroservoelastic state-space equations must be formulated in a reduced modal space, in order for the computational cost of the Lyapunov solver to remain tractable. Stresses (and their sensitivities) are known [12, 13] to be very difficult to accurately extract from modal approximations, as they converge slowly with mode number. Secondly, stress-based (or buckling-based) failure indices of interest, such as the von Mises stress, are nonlinear quadratic functions of the aeroservoelastic state, complicating the formulation of a stochastic failure criterion from the covariance matrix [14, 15].

These two complications have been addressed and solved in Refs [16] and [17], respectively, but neither has been entirely proven out in the context of a complete and realistic aeroservoelastic design exercise, which is the goal of the current work. This will be done for the Common Research Model, a highly flexible transport configuration with several electrohydraulic control surfaces distributed along the trailing edge.

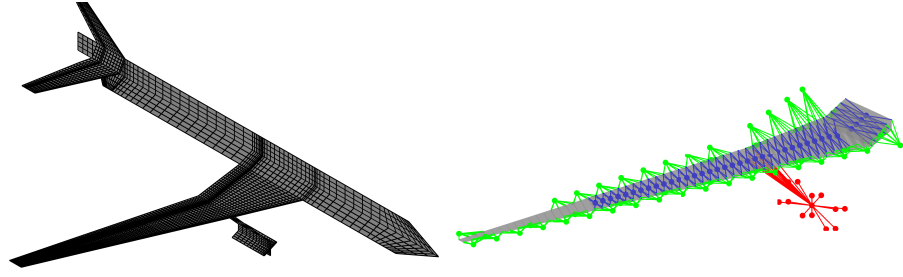
The optimization process, driven by adjoint derivatives, seeks to minimize structural wingbox mass via several different types of aeroservoelastic design variables. Detailed structural sizing variables govern the stiffness and inertial properties of the wingbox. Steady control surface deflections are optimized to modulate static maneuver loads, while unsteady control surface deflections (optimized via control law parameters) alleviate the dynamic gust response. Constraints are placed upon the static stress and buckling margins during maneuver loads, as well as the corresponding gust stress and buckling margins (computed with the strategies discussed above). The aeroelastic designs are obviously required to be free of flutter, but AFS is not included in this work due to the higher level of risk associated with this technology [18]. The effects of various constraints and design variables are sequentially investigated in this work, in order to provide a clear understanding of the effects of each upon the optimal aeroservoelastic performance.

2. Transport Configuration Test Case

All of the work in this paper is conducted on a conceptual high aspect ratio Common Research Model (CRM) jig shape shown in Fig. 1. This configuration, described extensively in Ref. [19], has an aspect ratio of 13.5, a wing span of 72 m, a mean aerodynamic chord of 6.3 m, a taper ratio of 0.25, a sweep angle of 35° , and a cruise Mach number of 0.85. The topology of the wingbox in Fig. 1 consists of 58 ribs, leading and trailing spars, and upper and lower surface skins. All shell members are outfitted with T-shaped stiffeners, where the flange is bonded to the shell members. The stiffeners are not modeled explicitly, but instead smeared into the shell stiffness properties [20]. An aluminum finite element model of this wingbox is composed of 21,300 triangular finite elements. Non-modeled secondary weight (of fasteners, etc.) is captured by increasing the structural density by a factor of 1.25. Nodes along the wing centerline are given a symmetric boundary condition; nodes along the wing root are given a pinned boundary condition, in order to approximate the wing/fuselage intersection.

Twenty articulated control surfaces are distributed along the trailing edge

Figure 2: Aerodynamic paneling (left) and non-structural lumped mass distribution (right).



of the wing from root to tip, also seen in Fig. 1. This is a departure from conventional aircraft layouts, whose trailing edge is populated with a sparser set of flaps, ailerons, and spoilers. This common layout may be considered a subset of the layout used here, though the current work also ignores any practical difficulties and limitations associated with simultaneous control of all twenty actuators. In particular, the use of several articulated control surfaces increases the probability of individual control failures, though may also be exploited to mitigate the shared risk.

For this work, the mass of each electrohydraulic actuator used to drive each control surface is set based on an initial estimate of the hinge moments encountered during maneuver loading, using a constant mass-per-hinge-moment value. This actuator mass value ranges from 214.5 kg (root) to 26.3 kg (tip). The mass of each control surface flap is proportional to the surface area of the wedge, and ranges from 140.4 kg (root) to 35.4 kg (tip). Actuators are modeled with third-order transfer functions, and are assumed to be irreversible, and as such the control deflections are unaffected by the aerodynamic hinge moments, simplifying the aeroservoelastic equations (detailed below) considerably [21].

The paneling scheme used to model steady and unsteady linear aerodynamic loads over the vehicle is shown in Fig. 2, with a total of 4,150 panels. A finer mesh is used toward the hinge line of the control surfaces, in order to improve the accuracy of the aeroservoelastic loads generated by deflections. Non-structural lumped mass distributions are also shown in the figure, to model the fuel (44,800 kg), engine (11,400 kg), and various unmodeled components in the leading and trailing edges of the wing (5,480 kg). The total mass of the 20 control surfaces and their actuators is 3,935 kg, and a fixed mass of 63,700 kg is used to model the remainder of the half-airplane (fuselage, etc.). Typical structural masses (for a single wingbox) will range during optimization from 10,000 to 20,000 kg. The TOGW for the entire vehicle is then roughly 300,000 kg.

3. Static Aeroelastic Modeling

Airplane maneuvers are considered to occur slow enough to warrant only static aeroelastic modeling: dynamic aeroelasticity is ignored in this section. Maneuver loads are computed with the following coupled linear static equation:

$$\mathbf{A}_s \cdot \mathbf{x}_s + \mathbf{B}_s \cdot \boldsymbol{\theta} = \mathbf{F}_s \quad (1)$$

This equation couples the finite element analysis, aerodynamic analysis, and the trim equations into a monolithic static aeroelastic analysis [22]. The static aeroelastic solution vector is $\mathbf{x}_s$, which contains the structural wingbox deformations, $\mathbf{u}$, the aerodynamic pressures, $\mathbf{C}_p$, the trim angle of attack, α , and the trim elevator setting δ . The matrix $\mathbf{A}_s$ contains the finite element stiffness matrix, the aerodynamic influence coefficients, and various load-displacement transfer vectors. The vector $\boldsymbol{\theta}$ is the commanded steady rotation of each trailing edge control surface, and $\mathbf{B}_s$ transfers the resulting loads onto the aeroelastic system. Finally, $\mathbf{F}_s$ contains vectors of inertial and thrust forces, jig shape downwash, and trim settings (desired load factor, etc.).

Unlike the dynamic aeroelastic equations that follow in the next section, the static structural deformation, $\mathbf{u}$, computed in Eq. 1 is done without the aid of a modal basis approximation, and is solved in the full finite element space. Having computed this deformation, plane stresses are computed for each shell finite element:

$$\left\{ \begin{array}{c} \sigma_x \\ \sigma_y \\ \sigma_{xy} \end{array} \right\}_e = \mathbf{B}_{\sigma_e} \cdot \mathbf{u} \quad (2)$$

where the matrix $\mathbf{B}_{\sigma_e}$ is different for each element e . To facilitate a buckling analysis, in-plane tractions may be computed for each stiffened skin panel (delineated by ribs):

$$\left\{ \begin{array}{c} N_x \\ N_{xy} \end{array} \right\}_e = \mathbf{B}_{N_e} \cdot \mathbf{u} \quad (3)$$

These panel-level tractions are computed via $\mathbf{B}_{N_e}$, which averages the stress values across the skin panel, and multiplies by the appropriate thickness value. Global buckling of each stiffened panel (bordered by ribs and spars) and local buckling between each stiffener are separately accounted for, via assumed simply supported boundary conditions along the perimeter of the panel, and an assumption of linearity.

Quadratic failure surfaces are given by $\mathbf{s}^T \cdot \mathbf{F} \cdot \mathbf{s} + \mathbf{f} \cdot \mathbf{s} \leq 1$. For stresses, the von Mises failure criterion is:

$$\mathbf{F} = \frac{1}{\sigma_Y^2} \begin{bmatrix} 1 & -1/2 & 0 \\ -1/2 & 1 & 0 \\ 0 & 0 & 3 \end{bmatrix} \quad \mathbf{f} = \mathbf{0} \quad (4)$$

For buckling, these terms are:

$$\mathbf{F} = \frac{1}{N_{xy,cr}^2} \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \quad \mathbf{f} = \frac{1}{N_{x,cr}} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad (5)$$

where the critical buckling loads $N_{x,cr}$ and $N_{xy,cr}$ are dependent upon the panel geometry, material properties, thickness, and smeared stiffener properties. Expressions for both local and global critical buckling envelopes are given in Ref. [20], along with other relevant details concerning the panel buckling computations used here.

The von Mises failure index is computed of each of the 21,300 finite elements via Eqs. 2 and 4, and then compressed into a set of 13 Kreisselmeier-Steinhauser (KS) functions [23]: three for the upper skins, three for the lower skins, three for the ribs, and four for the spars. Similarly, the global and local buckling failure indices are computed for each of the 114 upper and lower skin panels via Eqs. 3 and 5, and then compressed into 8 KS functions: four each for the upper and lower skins. If each of these KS functions is less than one, then the inequality $\mathbf{s}^T \cdot \mathbf{F} \cdot \mathbf{s} + \mathbf{f} \cdot \mathbf{s} \leq 1$ holds.

Six load cases are utilized here, summarized in Table 1, each of which is a longitudinal load case. Detailed aircraft design will of course be driven by a far larger set of cases (including a lateral maneuver such as a roll), but this set is reduced here in order to keep the computational cost tractable. Furthermore, no load case has been specifically dedicated to computing control surface reversal, though the closed-loop MLA load cases should provide information about poorly-performing control surfaces, and also provide the optimizer with information about how to improve the situation (i.e., structural stiffening), if needed.

Referencing Table 1, load cases 1 and 2 are both run at 2.5g, but case 1 features maneuver load alleviation from the control surfaces ($\boldsymbol{\theta}$ in Eq. 1), whereas case 2 does not: $\boldsymbol{\theta}$ is set to 0. Forcing the optimizer to account for load

Table 1: Summary of static aeroelastic load cases.

Case	N	Mach	Altitude	Safety Factor	Actuator Status
1	2.5	0.64	0 ft	1.5	closed-loop
2	2.5	0.64	0 ft	1.3	open-loop
3	-1	0.64	0 ft	1.5	closed-loop
4	-1	0.64	0 ft	1.3	open-loop
5	1	0.85	35,000 ft	1.5	open-loop
6	2	-	-	1.5	open-loop

cases that do not benefit from control surface-based aeroelastic tailoring (albeit at a lower safety factor) prevents overly-large mass reductions and injects a level of risk-reduction into the design process. In the unlikely event that the control surfaces are not working properly during flight, satisfaction of these open-loop load cases will ensure that the wing structure does not fail, though at a lower safety factor. A similar open/closed-loop dynamic is used for load cases 3 and 4, at $-1g$.

Load case 5 is a $1g$ cruise case, only included to facilitate gust loads in the following sections. The final load case 6 is a simple $2g$ inertial-only landing load, and is not aeroelastically coupled. Load case 6 is expected to dissuade the optimizer from allocating excessive material towards the wing tip (which otherwise may be beneficial from an inertial relief standpoint), and also size the lower skins for static buckling, along with the $-1g$ cases.

4. Dynamic Aeroelastic Modeling

The dynamic aeroelastic behavior of the wing is modeled in the time domain (state-space), via modal coordinates: $\mathbf{u} = \boldsymbol{\phi} \cdot \boldsymbol{\eta}$, where $\boldsymbol{\phi}$ is a matrix of mode shapes, and $\boldsymbol{\eta}$ is a vector of modal amplitudes. Rigid body modes are included, but only symmetric motion: hence the first two modes of $\boldsymbol{\phi}$ are pitch and plunge. The equations of motion are:

$$\dot{\mathbf{x}} = \mathbf{A} \cdot \mathbf{x} + \mathbf{B}_\theta \cdot \boldsymbol{\theta} + \mathbf{B}_\delta \cdot \delta + \mathbf{B}_w \cdot w \quad (6)$$

The dynamic aeroelastic solution vector $\mathbf{x}$ contains the modal amplitudes, $\boldsymbol{\eta}$, their time derivatives, $\dot{\boldsymbol{\eta}}$, aerodynamic lag states via a Roger approximation [24], control surface aerodynamic terms and their lag states, third-order actuator modeling terms, gust column aerodynamic lag terms, and gust forming filter terms. The control surface deflection vector $\boldsymbol{\theta}$ is retained from the previous section (though the actual control surface deflection experienced by the wing will be attenuated via the third-order actuator models), as is the elevator deflection δ . Finally, w is a white noise input meant to emulate a gust signal, which is sent through a forming filter embedded within $\mathbf{A}$. Additional details concerning Eq. 6 can be found in Refs. [7], [11], and [16].

A simple FCS is implemented via the equation $\delta = -\mathbf{K}_\delta \cdot \mathbf{x}$, which drives the elevator motion based on proportional control from the vehicle rigid body pitch and pitch rate. The terms within $\mathbf{K}_\delta$ are held fixed throughout all of the optimization results presented here.

4.1 Flutter Constraints

For the derivation of flutter constraints, AFS is not considered (i.e., $\boldsymbol{\theta} = \mathbf{0}$). Eigenvalues of $\bar{\mathbf{A}} = \mathbf{A} - \mathbf{B}_\delta \cdot \mathbf{K}_\delta$ are given by $s = \gamma + i \cdot \omega$. For a given Mach number, these eigenvalues may be tracked across a range of matched point equivalent air speeds. Speeds at which a given eigenvalue crosses into the right-half plane ($\gamma = 0$) are instability points: termed flutter if ω is non-zero, divergence in the case of $\omega = 0$. An instability constraint is formulated by imposing upper limits on the damping γ of each mode (either flutter or

divergence), at all equivalent air speeds (U_{EAS}) of interest [25]. Specifically, each damping value is constrained to lie beneath the piecewise polynomial:

$$\gamma \leq \begin{cases} 0 & 0 \leq U_{EAS} \leq U^* \\ c \cdot (U_{EAS} - U^*)^2 & U_{EAS} > U^* \end{cases} \quad (7)$$

The minimum allowable flutter equivalent air speed is U^* , computed from the desired flutter margin (1.15) and the flight envelope (using a 185 m/s dive speed). Critical U_{EAS} points (local minima) of the inequality in Eq. 7 are computed and lumped together into a single KS constraint. As above, if this constraint is greater than one, then Eq. 7 is not satisfied, and the structure does not meet the required flutter margin.

The quadratic parameter c is included in order to allow for the optimizer to “see” flutter mechanisms that are non-critical (i.e., a U_{EAS} larger than U^*), but may become critical across changes in the parameter space. The parameter, set to 500 for this work, must be large enough so as not to unduly penalize non-critical flutter mechanisms, but also small enough so that the design derivative of those non-critical flutter points are well-defined when they are in the vicinity of U^* .

4.2 Gust Constraints

The state covariance matrix, $\mathbf{X}$, due to a gust input, is computed via a Lyapunov equation:

$$\overline{\mathbf{A}} \cdot \mathbf{X} + \mathbf{X} \cdot \overline{\mathbf{A}}^T + \overline{\mathbf{Q}} = \mathbf{0} \quad (8)$$

where $\overline{\mathbf{Q}} = \mathbf{B}_w \cdot U_\sigma^2 \cdot \mathbf{B}_w^T$. The continuous turbulence design gust velocity, U_σ , is taken here to be 25.9 m/s [17, 26], and the gust is considered frozen in that it is unaffected by the presence of the airplane. Having computed the covariance $\mathbf{X}$, it is desired to compute the 3×3 stress covariance matrix (in terms of σ_x , σ_y , and σ_{xy}) for each finite element, and the 2×2 traction covariance matrix (in terms of N_x and N_{xy}) for each stiffened panel. Attempting to compute these via a transformation based upon $\mathbf{B}_{\sigma_e} \cdot \boldsymbol{\phi} \cdot \boldsymbol{\eta}$ will not work, since stresses are not easily extracted from modal quantities. Instead, modal acceleration (MA) must be used, a technique exhaustively described in Ref. [16], in the context of gust loads. Only cursory details will be given here.

The modally-reduced elastic, inertial, and aerodynamic terms that constitute the aeroservoelastic equations of motion in Eq. 6 can be alternatively written in the Laplace domain:

$$(s^2 \cdot \overline{\mathbf{M}}^{ss} + s \cdot \overline{\mathbf{C}}^{ss} + \overline{\mathbf{K}}^{ss} - q \cdot \overline{\mathbf{A}}^{ss}) \cdot \boldsymbol{\eta} + (s^2 \cdot \overline{\mathbf{M}}^{sc} - q \cdot \overline{\mathbf{A}}^{sc}) \cdot \boldsymbol{\Theta} - q \cdot \overline{\mathbf{A}}^{se} \cdot \delta = q \cdot \overline{\mathbf{A}}^{sg} \cdot w_g / U \quad (9)$$

where $\mathbf{M}$, $\mathbf{C}$, $\mathbf{K}$, and $\mathbf{A}$ are the mass, damping, stiffness, and aerodynamic matrices, respectively, and the overbar designates a modally-reduced quantity. q is the dynamic pressure, $\boldsymbol{\Theta}$ is the control surface rotation (related to $\boldsymbol{\theta}$ via a series of third order actuator transfer functions), and w_g is the speed of the gust column, related to w via a forming filter. The ss , sc , se , and sg superscripts designate structure-on-structure, control-on-structure, elevator-on-structure, and gust-on-structure partitioning, where the “structure” designation includes both rigid body modes and elastic modes.

For MA, the modally-reduced matrices in Eq. 9 must be *partially* un-projected back into the full-order finite element space. The columns of these matrices can remain in modal space, but the rows must be in the full-order finite element space. These new matrices may be designated as $\mathbf{M}^{ss} \cdot \boldsymbol{\phi}$, etc., where $\mathbf{M}^{ss}$ is the full-order mass matrix. The full-order structural deformation vector can

then be computed as:

$$\mathbf{u} = (\mathbf{K}^{ss})^{-1} \cdot \left(-(s^2 \cdot \mathbf{M}^{ss} + s \cdot \mathbf{C}^{ss} - q \cdot \mathbf{A}^{ss}) \cdot \boldsymbol{\phi} \cdot \boldsymbol{\eta} - (s^2 \cdot \mathbf{M}^{sc} - q \cdot \mathbf{A}^{sc}) \cdot \boldsymbol{\Theta} + q \cdot \bar{\mathbf{A}}^{se} \cdot \delta + q \cdot \mathbf{A}^{sg} \cdot w_g/U \right) \quad (10)$$

Converting Eq. 10 back from the Laplace domain into the state-space terms used for Eq. 6, leads to a MA equation of the form:

$$\mathbf{u} = \mathbf{H} \cdot \mathbf{x} \quad (11)$$

where $\mathbf{H}$ involves solving a system of equations of the full-order stiffness matrix $\mathbf{K}^{ss}$, with the number of right-hand-sides equal to the size of the aeroservoelastic solution vector $\mathbf{x}$. The 3×3 stress covariance matrix of each finite element is then given by:

$$\mathbf{C}_{\sigma_e} = \mathbf{B}_{\sigma_e} \cdot \mathbf{H} \cdot \mathbf{X} \cdot \mathbf{H}^T \cdot \mathbf{B}_{\sigma_e}^T \quad (12)$$

A similar 2×2 traction covariance matrix can be computed for the buckling of each stiffened panel.

Stresses are assumed to be the sum of $1g$ loads (computed via load case 5 in Table 1) and the gust loads:

$$\begin{Bmatrix} \sigma_x \\ \sigma_y \\ \sigma_{xy} \end{Bmatrix}_e = \mathbf{s}_e = \mathbf{s}_{e_1} + \mathbf{s}_{e_g} \quad (13)$$

An equal-probability hyper-surface, of the stochastic gust loads, is then written as:

$$\mathbf{s}_{e_g}^T \cdot (\mathbf{C}_{\sigma_e})^{-1} \cdot \mathbf{s}_{e_g} = 1 \quad (14)$$

If the surface of Eq. 14 lies entirely within the failure envelope of Eq. 4, then the gust loads are considered acceptable. The critical point along the hyper-surface is that with the closest proximity to the failure surface. This critical point can be found with a method developed in Ref. [17], and involves solving the following constrained optimization problem:

$$\begin{aligned} &\text{maximize: } \bar{g}_e = \mathbf{s}_e^T \cdot \mathbf{F} \cdot \mathbf{s}_e + \mathbf{f} \cdot \mathbf{s}_e - 1 \\ &\text{subject to: } (\mathbf{s}_e - \mathbf{s}_{e_1})^T \cdot (\mathbf{C}_{\sigma_e})^{-1} \cdot (\mathbf{s}_e - \mathbf{s}_{e_1}) = 1 \end{aligned} \quad (15)$$

If the optimal $\bar{g}_e$ value is less than 0, then the gust-stress constraint is satisfied. A graphical example of this is shown in Fig. 3 for a two-dimensional case.

$\bar{g}_e$ is computed for each shell finite element along the wingbox, and then compressed into a single KS aggregation function. Static stress values in the previous section had been compressed into 13 separate KS functions, but due to the large computational costs of computing the adjoint design derivatives of the MA matrix, $\mathbf{H}$ (discussed in more detail below), only a single aggregation is considered here. This choice adds more conservatism into the KS approximation.

Similar methods to Eqs. 12-15 may be considered for skin buckling, and again wrapped into a single KS constraint. A significant approximation is made here, however, in that the buckling envelope ($N_{x,cr}$ and $N_{xy,cr}$) used to build $\mathbf{F}$ and $\mathbf{f}$ in Eq. 15 is assumed to be static, independent of the loading frequency. This is known to be untrue [27], but the inclusion of a load-dependent failure envelope is not easily accommodated in the above framework, and is left for future work. Fatigue life, another potentially important design consideration for the gust-constrained problem considered here, is similarly ignored.

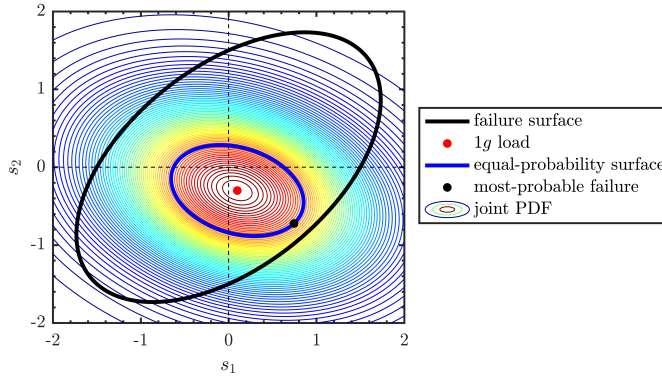


Figure 3: Two-dimensional example of an equal-probability hyper-surface and an elliptic failure boundary.

4.2.1 Open-Loop Gust Constraints

For open-loop analysis, the control surface rotations (commanded rotation θ and actual rotation Θ) are set equal to 0. The FCS controller driving the elevator motion δ remains active, and the safety factor associated with gust-based stress and buckling constraints is set to 1.5.

4.2.2 Closed-Loop Gust Constraints

For closed-loop analysis, both linear quadratic regulator (LQR) and linear quadratic Gaussian (LQG) controllers are used to drive the trailing edge control surfaces in this work. For LQR, under the assumption of stationary control and a continuous-time system, a controller is assumed of the form $\theta = -K_\theta \cdot x$, and computed by minimizing the typical cost function [28]:

$$J = \frac{1}{2} \cdot \int_0^\infty (x^T \cdot Q_x \cdot x + \theta^T \cdot R_\theta \cdot \theta) \cdot dt \quad (16)$$

where Q_x is the state weighting matrix, and R_θ is the control weighting matrix. The gust constraint process described above is the same, with the exception that now Eq. 8 uses $\bar{A} = A - B_\delta \cdot K_\delta - B_\theta \cdot K_\theta$. The MA matrix, H , is identical to the previous section, since the state vector x is independent of the *commanded* control surface rotations θ , but only a function of the *actual* control surface rotations Θ .

LQR assumes full-state feedback, which is rarely available. LQG assumes the presence of an output based on actual sensors: $y = C_x \cdot x + D_\theta \cdot \theta + D_\delta \cdot \delta + v$, where v is the sensor noise. For this work, six triaxial accelerometers are distributed along the wingbox, six strain gage rosettes are distributed along the upper skins, and six rosettes along the lower skins. The success of aeroservoelastic control is known to be dependent on optimal sensor placement [29], but such a study is out of scope for this work. Here, each of these sensor families (accelerometers, upper skin gages, lower skin gages) are allocated along the wingbox in a simple matrix of two chordwise and three spanwise.

The relevant $\bar{A}$ matrix is now augmented by observer terms:

$$\bar{A} = \begin{bmatrix} A - B_\delta \cdot K_\delta - B_\theta \cdot K_\theta & B_\delta \cdot K_\delta + B_\theta \cdot K_\theta \\ 0 & A - L \cdot C_x \end{bmatrix} \quad (17)$$

where L is the Kalman gain, computed with a similar cost function as in Eq. 16, but with the state weighting Q_x replaced with the process noise matrix $B_w \cdot U_\sigma^2 \cdot B_w^T$ from Eq. 8, and the control weighting R_θ replaced with the covariance matrix of the sensor noise v , designated as R_v [28]. The relevant $\bar{Q}$ matrix is now:

$$\bar{Q} = \begin{bmatrix} B_w & 0 \\ B_w & -L \end{bmatrix} \cdot \begin{bmatrix} U_\sigma^2 & 0 \\ 0 & R_v \end{bmatrix} \cdot \begin{bmatrix} B_w & 0 \\ B_w & -L \end{bmatrix}^T \quad (18)$$

Table 2: Optimization definition.

minimize:	structural mass	
subject to:	design variable side limits	$\mathbf{q}_{LB} \leq \mathbf{q} \leq \mathbf{q}_{UB}$
	static stress constraints	$KS_{\sigma,i} \leq 1$
	static buckling constraints	$KS_{\mu,i} \leq 1$
	flutter constraint	$KS_{fl} \leq 1$
	open-loop gust stress constraint	$KS_{\sigma_{gust,OL}} \leq 1$
	open-loop gust buckling constraint	$KS_{\mu_{gust,OL}} \leq 1$
	closed-loop gust stress constraint	$KS_{\sigma_{gust,CL}} \leq 1$
	closed-loop gust buckling constraint	$KS_{\mu_{gust,CL}} \leq 1$
	closed-loop gust rotation RMS constraint	$KS_{\dot{\theta}_{gust,CL}} \leq 1$
	stiffener - shell thickness difference limits	$KS_{t/t_s} \leq 1$
	stiffener aspect ratio limits	$KS_{AR} \leq 1$

Using these new definitions of $\bar{\mathbf{A}}$ and $\bar{\mathbf{Q}}$ in Eqs. 17 and 18, the state covariance matrix $\mathbf{X}$ may be computed via Eq. 8, where $\mathbf{X}$ has twice the number of states as for the LQR case, or the un-controlled case, due to the observer terms. These observer terms have no impact on the modal acceleration process, and so the MA matrix must simply be augmented by a zero-matrix: $\begin{bmatrix} \mathbf{H} & \mathbf{0} \end{bmatrix}$. The remaining MA process of Eqs. 12-15 is the same as above.

If GLA is to be considered in the optimization process, closed-loop gust constraints (stress and buckling) are augmented with a safety factor of 1.5. Similar to the methods used for static MLA, an open-loop gust constraint is retained, albeit at a lower safety factor of 1.3.

5. Optimization Definition

The optimization problem definition is given in Table 2, and is solved with a gradient-based method. The objective function to be minimized is the structural wingbox mass. Design variables $\mathbf{q}$ include sizing variables, MLA rotation variables, and control law variables. For sizing, the wingbox is broken into 283 design patches, where each rib-delineated skin panel is a patch, each rib-delineated spar section is a patch, and each rib is a patch. Within each patch, the thickness of each shell member (t , ranging from 3 to 30 mm), the thickness of the stiffeners attached to these shells (t_s , ranging from 2.5 to 30 mm), and the height of the stiffeners (h_s , ranging from 30 to 100 mm, except in the rib and spar webs, where the lower bound is 64 mm), are independently optimized. For this work, the height of each T-stiffener web is set equal to the flange width, and the thickness of the web and flange are equal as well, as drawn in Fig. 4. It should be emphasized that the aerodynamic jig shape of the wing is frozen here (planform, airfoil, etc.), as is the structural topology (layout of ribs and stiffeners): the sizing variables described above have no impact on these parameters.

MLA rotation design variables include the vector of commanded control surface rotations, $\boldsymbol{\theta}$, for the closed-loop maneuver load cases 1 and 3 in Table 1, bounded by $\pm 10^\circ$. These bounds are chosen to be large enough such that the MLA variables can appreciably impact the aerodynamic loads, but not so large that the optimizer will push the local flow state to a region that would not be well-modeled here (i.e., flow separation). The control law design variables are the members of the diagonal control weighting matrix $\mathbf{R}_\theta$ (one variable per control surface), bounded between 10^{-2} and 10^{12} . These final bounds are simply chosen to be wide enough such that they will be inactive throughout the optimization process.

Design constraints are also listed in Table 2, and include the static stress constraints KS_σ . The number of static stress constraints is $N_\sigma \cdot N_L$, where N_σ is the number of aggregation parameters (set to 13 as noted above), and N_L is the number of load cases (6 in Table 1). Similarly, KS_μ are the static buckling

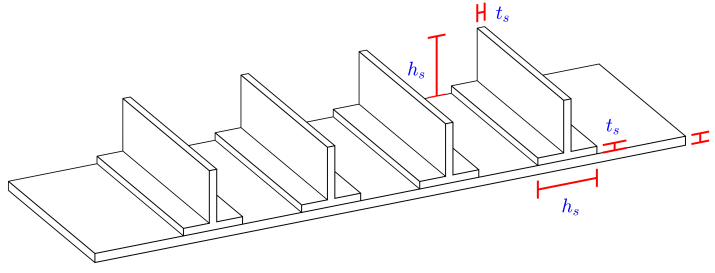


Figure 4: Example T-stiffener geometry.

constraints, numbering $N_\sigma \cdot N_\mu$ total, with N_μ set to 8. A flutter constraint (Eq. 7), open- and closed-loop stress-based gust constraints, and open- and closed-loop buckling-based gust constraints (Eq. 15) are also included. As a surrogate on controller cost, the RMS rotation rate for each control surface during closed-loop GLA is computed via the covariance matrix $\mathbf{X}$, aggregated into a single KS constraint $KS_{\dot{\theta}_{gust,CL}}$, and forced to be less than some threshold [2]. Ref. [30] provides guidance on reasonable ratios between rate saturation and the allowable RMS value (i.e., three standard-deviations): a range of limits are explored here. Furthermore, the same rotation rate limits are imposed on each control surface, an over-simplification given the range of actuator sizes used from root to tip.

The final two constraints in Table 2 regulate the stiffener geometry. It is desired that, for a given design patch, the shell thickness be within 2.5 mm of the stiffener thickness. These constraints are normalized to 1 for each design patch, and then all aggregated into a single metric, KS_{t/t_s} . Similarly, the aspect ratio of each stiffener should not be greater than 15, constraints which are normalized to 1 and then compressed into a single constraint, KS_{AR} .

Each set of design variables is passed through a linearly-decaying cone-shape filter [31] in order to prevent the difference in stiffness between adjacent patches, or the difference in actuation between adjacent control surfaces, from being too large. The gradient-based optimization problem of Table 2 is solved with the Globally-Convergent Method of Moving Asymptotes tool (GCMMA) [32]. Design derivatives of the static aeroelastic system in Eq. 1 are computed with the adjoint method, and derivatives of the flutter constraint in Eq. 7 are computed with eigenvalue sensitivity analysis methods [33]. Derivatives of the gust constraints are substantially more complex. Starting with Eq. 15, four design variable dependencies must be included in the chain rule:

- Derivatives of $\bar{g}_e$ with respect to the covariance matrix $\mathbf{X}$, and explicit derivatives of $\mathbf{X}$ with respect to the design variables $\mathbf{q}$. The latter involves differentiating the Lyapunov equation, and for closed-loop cases, derivatives of the LQR or LQG controllers must also be included [34].
- Derivatives of $\bar{g}_e$ with respect to the MA matrix $\mathbf{H}$, and explicit derivatives of $\mathbf{H}$ with respect to the design variables $\mathbf{q}$. The derivatives of $\mathbf{H}$ are computed via the adjoint method, which involves solving a system of equations on the stiffness matrix $\mathbf{K}^{ss}$, with the number of right-hand-sides equal to the size of the covariance matrix $\mathbf{X}$ multiplied by the number of KS constraints. Given the large cost of this operation, only a single KS constraint is included for stress and buckling in Table 2, as noted previously.
- Derivatives of $\bar{g}_e$ with respect to the $1g$ loads $\mathbf{s}_{e1}$, computed via an adjoint computation of Eq. 1.
- For buckling only, derivatives of $\bar{g}_e$ with respect to the failure envelope $\mathbf{F}$ and $\mathbf{f}$, and the derivative of this envelope with respect to the sizing design variables.

Further details of the gust constraint sensitivity analysis process are outlined in Ref. [16] and [17].

Figure 5: Optimization convergence with and without MLA.

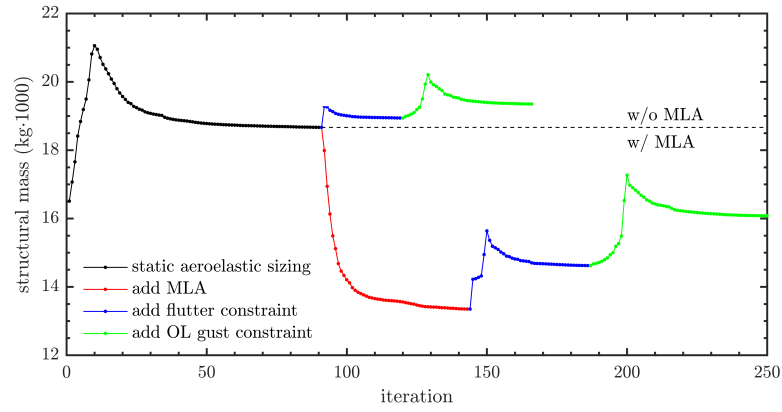
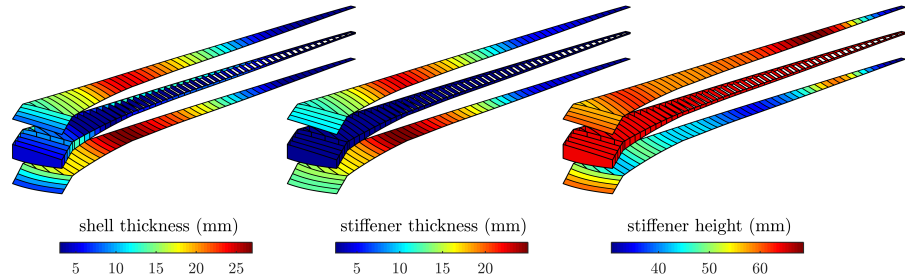


Figure 6: Optimization distribution of t , t_s , and h_s along the wingbox.



6. Results

Aeroelastic optimization results are conducted in five stages. First, only static aeroelastic optimization of an un-sized baseline wingbox is considered, without MLA. Secondly, MLA-based design variables (i.e., the vector of control surface rotations, θ) are included. Third, a flutter constraint is added. Fourth, open-loop gust constraints are added. Fifth, closed-loop gust constraints are added, as are the controller-based design variables (i.e., the members of the matrix $\mathbf{R}_\theta$). The convergence results of stages 1-4 are shown in Fig. 5, with and without MLA. Stage 1 converges to a structural wingbox mass of 18,666 kg; the optimal sizing parameters along the wingbox are shown in Fig. 6, in terms of t , t_s , and h_s of each design patch.

This first-stage aeroelastic sizing converges after 91 iterations; MLA design variables (lower portion of Fig. 5) are included between iterations 92 and 144. Structural mass decreases substantially, from 18,666 kg to 13,349 kg, and the optimal control surface rotation variables for the closed-loop cases 1 (2.5g) and 3 (−1g) are shown in Fig. 7. For the positive load case, MLA shifts the center of lift towards the root, with negative rotations (load alleviation) outboard for bending moment reductions, and positive rotations (load augmentation) inboard to help maintain trim. The optimal sizing variables are qualitatively similar to those seen in Fig. 6 (though thinner throughout), and are not shown. The mass reduction due to MLA is expected to be proportional to the open-loop safety factors, set to 1.3 here. A lower value will allow the optimizer to remove additional mass; a value of 1.5 (equal to that attached to the closed-loop cases) will revert the structural mass back to 18,666 kg, which is the same value found by the optimizer without MLA included.

After the second-stage MLA results converge at iteration 144, a flutter constraint is added (increasing the structural mass to 14,622 kg), and an open-loop gust constraint is subsequently added (increasing the structural mass to 16,078 kg). These structural mass penalties due to dynamic aeroelastic constraints are substantial: relative to the static aeroelastic MLA result, the flutter constraint penalizes mass by 9.5%, and the open-loop gust constraint by 20.4%. This is in strong contrast to the results without MLA included (upper portion of Fig. 5), where the flutter and gust constraints only add 1.4% and 3.6%, respectively.

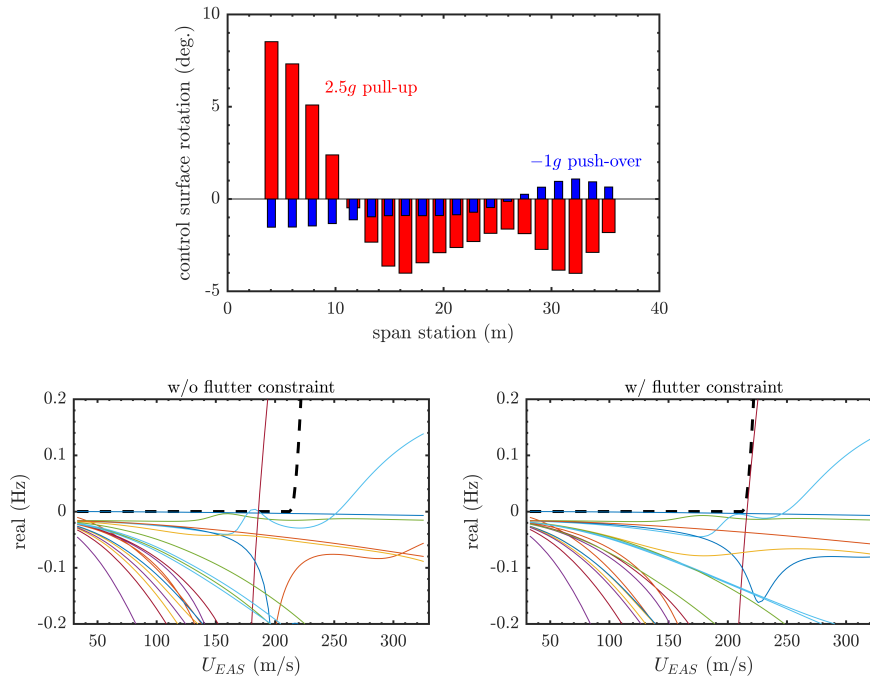


Figure 7: Optimal control surface rotations during the two closed-loop static aeroelastic maneuvers.

Figure 8: Eigenvalue migration (real part) before and after the flutter constraint is implemented; constraint boundary indicated by thick dashed line.

Only aeroelastic optimization with MLA included is considered for the remainder of the results in this paper.

Eigenvalue migration plots (real part only) of the third-stage design results are shown in Fig. 8 (real part) and Fig. 9 (imaginary part), both before and after the addition of the flutter constraint in Eq. 7. This flutter constraint is highly active for the final feasible design on the right side of Fig. 8, with four roots strongly interacting with the constraint boundary: the phugoid mode, the second bending mode (exhibits a “hard” flutter mechanism), an in-plane bending mode, and a torsional engine-pitch mode (both exhibiting softer “hump” flutter mechanisms). It should be noted that one or both of these hump mechanisms may become non-critical with the advent of structural damping, which has not been included in this work. Finally, the open-loop buckling constraint is added to Fig. 5 for the fourth design stage. Flutter plots for this final design are very similar to those shown in Fig. 8, and are not given here. Gust-based buckling constraints ($\bar{q}_e$ from Eq. 15) are shown in Fig. 10.

The gust buckling constraint is less than 0 for all of the upper and lower skin panels throughout the wingbox, indicative of a feasible design. The constraint is in fact active, but the largest $\bar{q}_e$ is -0.21, which is a result of the conservatism of the KS function. The equal-probability surface, failure surface, 1g loads, and the most-probable failure point (i.e., the point which maximizes $\bar{q}_e$ in Eq. 15) are also shown in the figure, for three selected panels. The axial loads N_x are clearly more prevalent than the shear loads N_{xy} for both the steady 1g case as well as the probabilistic gust covariance. Panel 66 (lower skins towards the

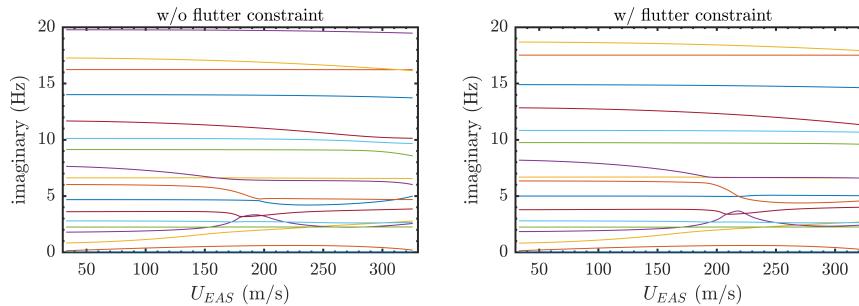


Figure 9: Eigenvalue migration (imaginary part) before and after the flutter constraint is implemented.

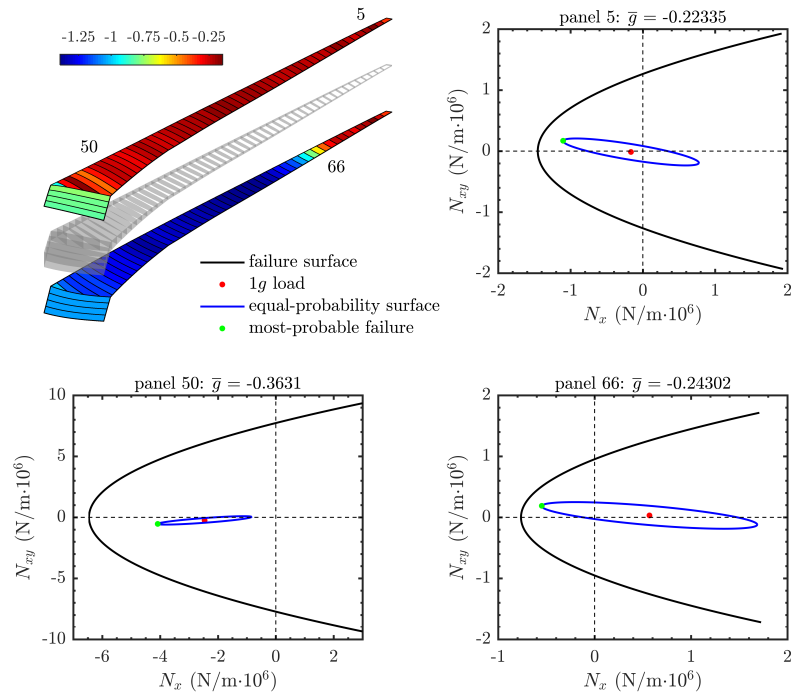


Figure 10: Open-loop gust-based buckling constraints along the paneled wingbox, and probabilistic failure boundaries for three selected panels.

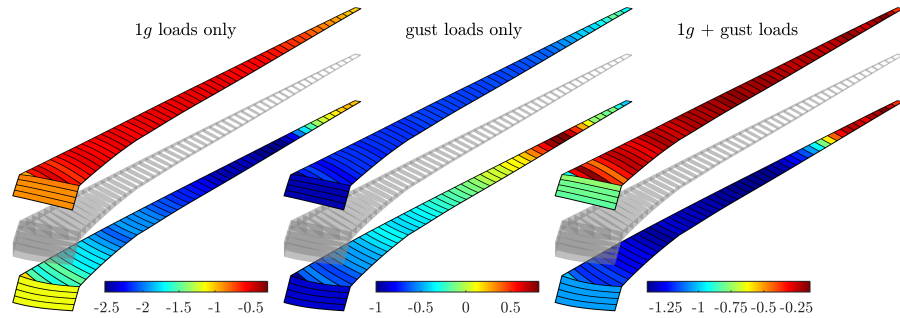


Figure 11: Open-loop gust-based buckling constraints along the paneled wingbox, with isolated contributions from trim loads (left) and gust loads (center).

wingtip) is a case where the gust loads alone would be large enough to cause a failure, but the stabilizing effect of the $1g$ loads (namely the tensile stresses along the lower skins of the deformed wing) retains the equal-probability surface within the failure envelope. This is in contrast to panels 5 and 50 in Fig. 10 (upper skins at tip and root, respectively), where the gust loads and the trim loads are largely additive. Finally, it is repeated here that the failure boundaries of Fig. 10 are assumed to be static, when in fact there will be some dependence on the loading frequency [27], rendering this failure surface non-deterministic in addition to the gust response.

Additional insight into these buckling constraints is given in Fig. 11, which shows isolated contributions to the $\bar{g}_e$ constraint from trimmed $1g$ loads only (by setting the covariance matrix to 0), from gust loads only (by setting the

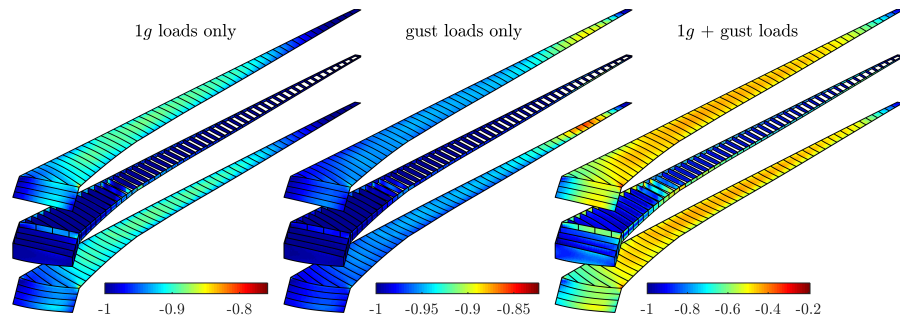


Figure 12: Open-loop gust-based stress constraints along the paneled wingbox, with isolated contributions from trim loads (left) and gust loads (center).

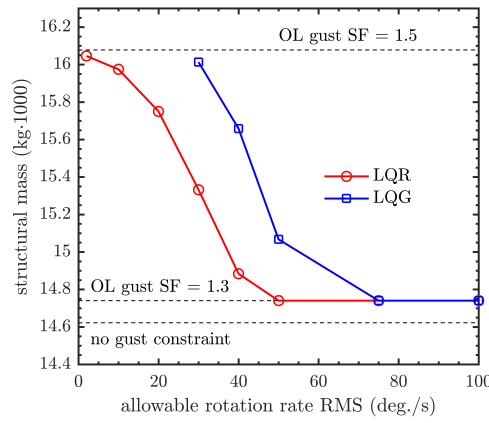


Figure 13: Structural mass increase with reduced controller effectiveness.

1g loads to 0), and the combined contribution. This final third plot in Fig. 11 of the combined contribution is reproduced from Fig. 10. Similar results are given for the gust-based stress constraints in Fig. 12. Stress-based failure and equal-probability surfaces for various points along the wingbox are not easily plotted (analogous to the results in Fig. 10), owing to the third dimension: both surfaces are three-dimensional ellipsoids.

The complete gust constraints in Figs. 11 and 12 are not a linear sum of the individual trim and gust contributions, owing to the highly nonlinear operators involved in the solution to Eq. 15. For buckling, the trimmed 1g constraint values are higher in the upper skins than lower, as the former is deformed in compression, and thus more buckling prone. Furthermore, these buckling constraints are nearly uniform along the upper surface, which is largely the result of the optimizer sizing each panel so as to remain free of buckling during the 2.5g static loads. Though not shown here, this 2.5g static buckling constraint is nearly active throughout the entire upper surface; the 1g buckling constraint is nearly uniform as well (though of course not active). Gust constraints purely due to the gust covariance (middle drawing in Figs. 11) are concentrated outboard, presumably due to the high inertial loads. This leads to the aforementioned situation where pure gust-based buckling loads are infeasibly large; only the addition of the stabilizing 1g loads produces a feasible combined response.

The dynamic stress constraints in Fig. 12 show similar trends to the buckling results: peak static 1g stresses occur at the stress concentration near the trailing edge of the wing root. Combining this result with the dynamic gust loads helps push the high stress constraints outboard. The aggregate stress constraint is active in the final design, but again owing to the conservatism of the KS method, the highest stress-based $\bar{g}_e$ for any of the finite elements is only -0.2. Strategies to adaptively reduce the conservatism of KS constraints have been addressed (see Ref. [35], for example), but are not pursued here.

Design stages 1-4 are shown in Fig. 5: 1) initial static aeroelastic sizing, 2) MLA, 3) flutter constraints, and 4) open-loop gust constraints. The final stage is the introduction of closed-loop gust constraints. This is done by reducing the safety factor of the open-loop gust constraints (stress and buckling) from 1.5 to 1.3, adding closed-loop gust constraints (stress and buckling) with a safety factor of 1.5, adding a constraint on the maximum allowable control surface rotation rate RMS [2], and finally adding the control weighting design variables (members of $\mathbf{R}_\theta$). The optimal structural wingbox mass, across a range of RMS rate limits, is shown in Fig. 13, for both LQR and LQG controllers.

In addition to the LQR/LQG data, three additional structural mass values are indicated in Fig. 13, to bracket the data. First is the structural mass without any gust constraints, 14,622 kg, also shown in Fig. 5 at the end of the flutter-constrained optimization stage-3. Second is the structural mass when only an open-loop gust constraint is included with a safety factor of 1.5: 16,078 kg, also

shown in Fig. 5 at the end of the gust-constrained stage-4. The prime goal here is to erase/recover this latter impact through the use of GLA (AFS could also be used to recover the mass penalty arising from the flutter constraint [18], but this is not considered here). The final mass value shown in Fig. 13 is the optimal structural mass when only an open-loop gust constraint with a safety factor of 1.3 is considered: 14,741 kg. It can then be seen that the mass penalty when imposing an open-loop 1.5 gust constraint is substantial, but only a minor change arises from an open-loop 1.3 gust constraint. This last structural mass value of 14,741 kg is relevant because the GLA optimization cases (both LQR and LQG) converge to it, when large control surface rotation rates are allowed. In other words, if control cost is of no concern, the GLA controller is able to satisfy the closed-loop stress and buckling constraints entirely, without the need to stiffen the structure (and thus increase the structural mass). The wingbox can only satisfy the open-loop gust constraints using passive means.

Decreasing the allowed rotation rate RMS (i.e., making this constraint harder to satisfy), generally reduces the effectiveness of the GLA controller, and forces the closed-loop gust constraints to be partially borne by the wingbox itself, increasing the structural mass. For LQR, this begins to happen when the allowable rate RMS is forced below 50 deg./s. For RMS values below 2 deg./s, the LQR controller is completely ineffective, and the closed-loop gust constraints must be entirely satisfied by the wingbox. The optimal structural mass regresses to the same value found when only an open-loop gust constraint is included with a safety factor of 1.5: 16,078 kg. In this limiting case, the closed-loop constraint has effectively been converted into an open-loop constraint.

It can also be seen in Fig. 13 that the LQG controller is less effective than its LQR counterpart. This is perhaps expected given that the LQG controller is driven by a finite amount of sensor data (triaxial accelerometers and strain gage rosettes), whereas LQR assumes full-state feedback. The LQR controller loses most of its effectiveness when the rotation rate RMS is forced below 2 deg./s, but this occurs for LQG at a much higher value of 30 deg./s. Of course, the LQG controller is defined by many parameters which have not been fully explored here, including sensor noise $\mathbf{R}_v$ and the number/location/type of the sensors themselves. Different values for these parameters may change the LQG performance seen in Fig. 13.

The control surface rotations are shown for both LQR and LQG controllers in Fig. 14, for the case where there is no constraint on rotation rate RMS. As noted above, in this case both controllers are able to entirely satisfy the closed-loop gust constraints without the need to increase structural mass. Rotation rates are quite high as a result, with peak RMS values reaching 120 deg./s for the LQR controller, and 168 deg./s for LQG. The energy distribution of the two controllers is very different as well, with LQR concentrating high rotation rates towards the root (presumably driven by rigid body motions), while LQG is more evenly distributed over the wing. Though not included in the optimization in any way, control surface rotation angle RMS is shown in Fig. 14 as well, with the general trends mirroring the angular rate RMS values [30], as expected. Elevator motions, driven by the FCS, are also shown in the figure, with the understanding that angular RMS values of the oscillating elevator would in reality be superimposed upon the steady elevator deflection needed for 1g trim (which is not shown).

Finally, an extension of Fig. 14 is shown in Fig. 15, for the controlled cases where the peak RMS rate is limited to 40 deg./s. As with previous results, the largest rate RMS value seen in the figure is less than 40 deg./s, owing to the conservatism of the KS constraint. Unlike with the previous un-constrained result in Fig. 14, the constrained results in Fig. 15 demonstrate a less-smooth variation in control surface motion along the trailing edge: this is due to the optimizer's aggressive use of the control law design variables ($\mathbf{R}_\theta$) in order to satisfy design constraints. These optimal values are not shown here, but can

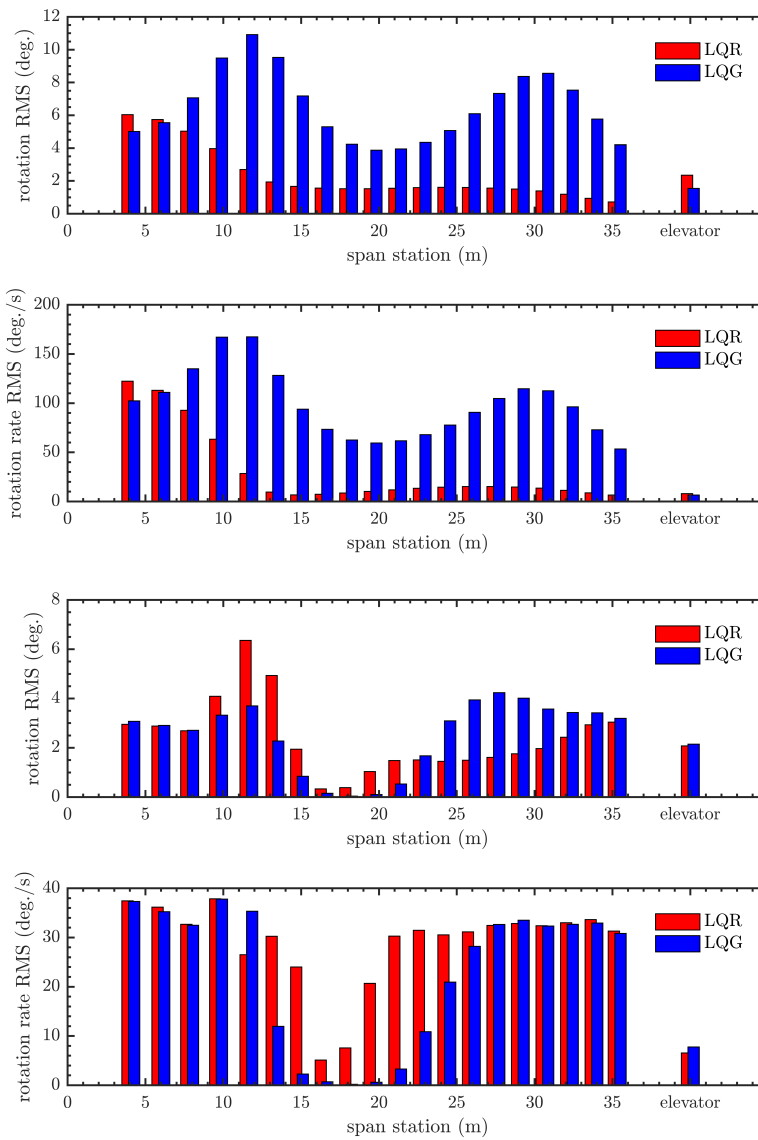


Figure 14: Control surface and elevator RMS motion for the case without limits on rate RMS.

Figure 15: Control surface and elevator RMS motion for the case where rate RMS is limited to 40 deg./s.

vary by orders of magnitude from the wing root to the tip.

7. Conclusions

This work has considered aeroservoelastic optimization of the Common Research Model transport configuration, with a standard rib/spar/skin/stiffener wingbox, and outfitted with several electrohydraulic control surfaces along the trailing edge. These control surfaces are used for both maneuver load alleviation and gust load alleviation. The objective function is to minimize structural mass, and design variables are set for detailed sizing of the wingbox members, steady control surface deflections during maneuver load alleviation, and distributed parameters used to define the closed-loop controllers for gust load alleviation. Aeroelastic design constraints include steady stress and buckling metrics across several trimmed maneuver loads, as well as flutter constraints, but the novel aspect of this paper is in the gust constraint formulation, which involves computing stress/buckling metrics in response to a stochastic continuous gust input. The formulation of these constraints echoes previous work done in Refs. [16] and [17], but is demonstrated here, for the first time, in the context of a complete design optimization for a relatively realistic airplane configuration.

A summary of the various cases considered in this work is tabulated in

Table 3: Summary of optimization cases (1).

case	MLA?	flutter margin	OL gust SF	CL gust SF	rate	RMS
1	no	-	-	-		-
2	no	1.15	-	-		-
3	no	1.15	1.5	-		-
4	yes	-	-	-		-
5	yes	1.15	-	-		-
6	yes	1.15	1.3	-		-
7	yes	1.15	1.5	-		-
8	yes	1.15	1.3	1.5	no limit, LQR	
9	yes	1.15	1.3	1.5	no limit, LQG	
10	yes	1.15	1.3	1.5	40 deg./s, LQR	
11	yes	1.15	1.3	1.5	40 deg./s, LQG	

Table 3 and 4, with design variables and constraints phased in sequentially. At each design phase, each of the available constraints is active. Adding a gust constraint to a design which is already flutter-constrained, for example, returns a design which is both gust-constrained and flutter-constrained. As such, the latter designs in the tables are highly-constrained, with both the open-loop and closed-loop gust constraints active. The sensitivity of the optimal results to the introduction of new design variables or constraints can be inferred from the subsequent change (a decrease for introduce design variables, an increase for introduced constraints) in the optimal structural mass.

Referencing Table 3 and 4, the dynamic aeroelastic constraints (flutter and gust) do not have a sizable impact on the problem if maneuver load alleviation is not considered, as the static stress and buckling constraints are stronger design drivers. Introducing maneuver load alleviation allows the optimizer to remove a significant amount of structural mass, thus rendering both flutter and gust as stronger drivers. If control cost is of no concern, the increase in structural mass introduced through the gust constraints can be largely recovered through closed-loop control. Handicapping the control surface rotation rates limits the effectiveness of the controllers, forcing the optimizer to add structural mass in order to (partially, at least) passively control the gust response. LQG controllers are more susceptible to this degradation than LQR, given that they are driven by a finite amount of sensor data.

The optimization process considered here has only considered linear aeroservoelastic models, though the configuration under consideration would likely be subject to a number of nonlinearities, including large structural deformations and aerodynamic shocks, which should be included in future work. If gust perturbations can be considered small relative to the trimmed cruise state, then a linearized analysis will suffice, and many of the gust-based design ideas implemented here (forming filters, Lyapunov equations, equal-probability failure surfaces, etc.) will convey; i.e., the nonlinearities will be confined to the static deformation.

The results presented here are of a preliminary nature, due to the above issue, as well as the fact that the number of load cases considered here is far fewer than would need to be considered in practice. However, the current result set does provide a groundwork for handling a set of certain aeroservoelastic metrics (i.e., open- and closed-loop stochastic gust constraints via detailed structural stress and buckling) that are not normally included in the design process, and further demonstrates the quantitative impact these constraints have on the optimization process.

case	structural mass	figure
1	18,666 kg	Fig. 6
2	18,939 kg	Fig. 5
3	19,349 kg	Fig. 5
4	13,349 kg	Fig. 7
5	14,622 kg	Fig. 8
6	14,741 kg	Fig. 13
7	16,078 kg	Fig. 10
8	14,741 kg	Fig. 14
9	14,741 kg	Fig. 14
10	14,883 kg	Fig. 15
11	15,660 kg	Fig. 15

Table 4: Summary of optimization cases (2).

Acknowledgments

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