

# Bi-Level Integrated System Synthesis (BLISS) 

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# BI-LEVEL INTEGRATED SYSTEM SYNTHESIS (BLISS) 

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#### Abstract

BLISS is a method for optimization of engineering systems by decomposition. It separates the system level optimization, having a relatively small number of design variables, from the potentially numerous subsystem optimizations that may each have a large number of local design variables. The subsystem optimizations are autonomous and may be conducted concurrently. Subsystem and system optimizations alternate, linked by sensitivity data, producing a design improvement in each iteration. Starting from a best guess initial design, the method improves that design in iterative cycles, each cycle comprised of two steps. In step one, the system level variables are frozen and the improvement is achieved by separate, concurrent, and autonomous optimizations in the local variable subdomains. In step two, further improvement is sought in the space of the system level variables. Optimum sensitivity data link the second step to the first. The method prototype was implemented using MATLAB and iSIGHT programming software and tested on a simplified, conceptual level supersonic business jet design, and a detailed design of an electronic device. Satisfactory convergence and favorable agreement with the benchmark results were observed. Modularity of the method is intended to fit the human organization and map well on the computing technology of concurrent processing.


## 0. Introduction

Optimization of complicated engineering systems by decomposition is motivated by the obvious need to dis-

[^0]tribute the work over many people and computers to enable simultaneous, multidisciplinary optimization. It is important to partition the large undertaking into subtasks, each small enough to be easily understood and controlled by people responsible for it. This implies granting people in charge of a subtask a measure of authority and autonomy in the subtask execution, and allowing human intervention in the entire optimization process.

Reconciliation of the need for subtask autonomy with the system level challenge of "everything influences everything else" is difficult. Each of the leading MDO methods that have evolved to date (survey papers: Balling and Sobieszczanski-Sobieski, 1996; and Sobi-eszczanski-Sobieski, J., and Haftka, R. T, 1997) tries to address that difficulty in a different way. In the system optimization based on the Global Sensitivity Equations (GSE) (Sobieszczanski-Sobieski, J. 1990, Olds, J. 1992, Olds, J. 1994), the partitioning applies only in the sensitivity analysis while optimization involves all the design variables simultaneously. The Concurrent SubSpace Optimization method provides for separate optimizations within the modules (SobieszczanskiSobieski, J. 1988, Renaud and Gabriele, 1991, 1993, and 1994; Stelmack and S. Batill, 1998) but handles all the design variables simultaneously in the coordination problem. The Collaborative Optimization method (Braun and Kroo, 1996; Sobieski and Kroo, 1998) also enables separate optimizations within the modules, each performed to minimize a difference between the state and design variables and their target values set in a coordination problem. This problem combines the system optimization with the system analysis, therefore its dimensionality may be quite large.

Most of the above method implementations had to overcome difficulties with integration of dissimilar codes. This has stimulated use of Neural Nets and Response Surfaces as means by which subdomains in the design space may be explored off-line and still be represented to the entire system. Unfortunately, effectiveness of this approach is limited to approximately 12 to 20 independent variables, hence, it is best suited for the early design phase. Consequently, a clear need remains
for a method applicable in later design phases when the number of design variables is much larger. Methods that build a path in design space fit that requirement. Ultimately, one needs both domain-exploring methods and path-building methods, enhanced with seamless 'gear-shifting' between the two.

Motivated by the above state of affairs, BLISS attacks the problem by performing an explicit system behavior and sensitivity analysis using the GSE, autonomous optimizations within the subsystems performed to minimize each module contribution to the system objective under the local constraints, and a coordination problem that engages only a relatively small number of the design variables that are shared by the modules. Solution of the coordination problem is guided by the derivatives of the behavior and local design variables with respect to the shared design variables. These derivatives may be computed in two different ways, giving rise to two versions of BLISS.

In either version, BLISS builds a gradient-guided path, alternating between the set of disjointed, modular design subspaces and the common system-level design space. Each segment of that path results in an improved design so that if one starts from a feasible state, the feasibility in each modular design subspace is preserved while the system objective is reduced. In case of an infeasible start, the constraint violations are reduced while the increase of the objective is minimized. Be cause the system analysis is performed at the outset of each segment of the path, the process can be terminated at any time, if the budget and time limitations so require, with the useful information validated by the last system analysis. In addition to enabling complete human control in the subspace optimization, BLISS allows the engineering team to exercise judgment, at any point in the procedure, to intervene before committing to the next successive pass.

BLISS has been developed in a prototype form and has been successfully demonstrated on the small-scale test cases reported herein.

## 1. Notation

BB - black box, a module, in the mathematical model of a system.
$\operatorname{BBA}\left(Y_{r},\left(Z, X_{r}\right)\right)$ - analysis of $B B_{r}$ to compute $Y_{r}$ for given $Z$ and $X_{r}$
$\mathrm{BBOF}_{\mathrm{r}}$ - BB Objective Function computed in $\mathrm{BB}_{\mathrm{r}}$ BBOPT $_{r}\left(\mathrm{X}_{\mathrm{r}}, \phi_{\mathrm{r}}, \mathrm{G}_{\mathrm{r}}\right)$ - optimization in $\mathrm{BB}_{\mathrm{r}}$ defined by eq. (2.1/9)
$\operatorname{BBOSA}_{\mathrm{r}}\left(\mathrm{X}_{\mathrm{r}, \mathrm{opt}}, \mathrm{Z}, \mathrm{Y}_{\mathrm{r}, \mathrm{s}}\right)$ - analysis of BB optimum for
sensitivity to parameters
$\operatorname{BBSA}\left(\mathrm{D}\left(\mathrm{Y}_{\mathrm{r}},\left(\mathrm{Z}, \mathrm{X}_{\mathrm{r}}, \mathrm{Y}_{\mathrm{r}, \mathrm{s}}\right)\right.\right.$ ) - sensitivity analysis of $\mathrm{BB}_{\mathrm{t}}$ to compute its output derivatives w.r.t. $\mathrm{Z}, \mathrm{X}_{\mathrm{r}}$, and $\mathrm{Y}_{\mathrm{r}, \mathrm{s}}$ $\mathrm{D}(\mathrm{V} 1, \mathrm{~V} 2)$ - total derivative $\mathrm{dV} 1 / \mathrm{dV} 2$
$\mathrm{d}(\mathrm{V} 1, \mathrm{~V} 2)$ - partial derivative $\partial \mathrm{V} 1 / \partial \mathrm{V} 2$;
D() , and d() dimensionality depends on the dimensionalities of V1 and V2:
V1 and V2 - are both scalars, then D and d are scalars
V1 vector, V2 scalar, then $D$ and $d$ are vectors V1 scalar, V2 vector, then D and d are vectors V1 vector, V2 vector, then D and $R$ are matrices
$G_{o}$ - vector of constraints active at the constrained minimum, length $\mathrm{NG}_{\mathrm{o}}$
$\mathrm{G}_{\mathrm{r}}$ - vector of the constraint functions, $\mathrm{g}_{\mathrm{r}, \mathrm{l}}$ local to $\mathrm{BB}_{\mathrm{r}}$, $\mathrm{g}_{\mathrm{r}, \mathrm{t}} \leq 0$ is a satisfied constraint
$\mathrm{G}_{\mathrm{yz}}$ - constraints in a BB that have a stronger depend ence on $Y$ and $Z$, than on $X$
GSE - Global Sensitivity Equations (SobieszczanskiSobieski, 1990); GSE/OS - GSE/Optimized Subsystems.
I - identity matrix.
L - vector of the Lagrange multipliers corresponding to $G_{o}$, length $\mathrm{NG}_{\mathrm{o}}$
LP - Linear Programming
NB - the number of BBs in the system
NLP - NonLinear Programming
opt - subscript denotes optimized quantity
P - vector of parameters, $p_{i}$, kept constant in the process of finding the constrained minimum, length NP.
SA((P,Z,X),Y) - system analysis; a computation that outputs $Y$ for a system defined by $P, Z$, and $X$
SOF - System Objective Function computed in one of the BBs
SOPT(Z,Ф) - system objective optimization defined by eq. (2.2.3/1)
$\operatorname{SSA}(\mathrm{D}(\mathrm{Y},(\mathrm{Z}, \mathrm{X}))$ - system sensitivity analysis to compute sensitivity of the system response Y w.r.t. Z and $X$
TOGW - take-off gross weight
T-superscript denotes transposition.
$X_{r}$ - vector of the design variables $X_{r, j}$, length $N X_{r}$, these variables are local to $\mathrm{BB}_{\mathrm{r}} ; \mathrm{X}$ without subscript - a vector of all concatenated $X_{r}$, length $N X$
$\mathrm{XL}, \mathrm{XU}$ - lower and upper bounds on X , sideconstraints.
$Y_{r}$ - vector of behavior (state) variables output from $\mathrm{BB}_{\mathrm{r}}$, these are the coupling variables; an element of $Y_{r}$ is denoted $y_{r, i}$; some of $y_{r, i}$ are routed as inputs to other BBs, and may also be routed as output to the outside; the $\mathrm{Y}_{\mathrm{r}}$ length is $\mathrm{NY}_{\mathrm{r}} ; \mathrm{Y}$ without subscripts - a vector of all concatenated $Y_{r}$, length NY
$Y_{r, s}$ - vector of variables input to $\mathrm{BB}_{\mathrm{r}}$ from $\mathrm{BB}_{\mathrm{s}}$, these are the coupling variables; an element of $\mathrm{Y}_{\mathrm{r}, \mathrm{S}}$ is denoted $y_{r, s, i}$; note that by this definition $Y_{r, s}$ is a subset of $Y_{r}$, vector length $N Y_{r, s}$
$Z$ - vector of the design variables $\mathrm{Z}_{\mathrm{k}}$ that are shared by two or more BBs, these are the system-level variables; length NZ
0 - subscript denotes the present state from which to extrapolate, or the optimal state.
$\mathrm{ZL}, \mathrm{ZU}$ - lower and upper bounds on Z , sideconstraints
$\Delta$ - increment
$\Delta \mathrm{ZL}, \Delta \mathrm{ZU}$ - move limits
$\phi_{\mathrm{r}}$ - the local objective function in $\mathrm{BB}_{\mathrm{r}}$
$\Phi$ - the system objective function equated to one, particular $y_{1, i}$

## 2. The Algorithm

In this section, the symbols defined in Notation are used in a shorthand manner without repeating their definitions.


Figure 1: System of Coupled BBs
The algorithm is introduced using an example of a generic system of three BBs, as shown in Figure 1. Three is a number small enough for easy conceptual grasp and compact mathematics, yet large enough to unfold patterns that readily generalize to larger NB. Even though the system in Figure 1 is generic, it may be useful to bear a specific example in mind. Let it be an aircraft so that:

BB1 - performance analysis
BB2 - aerodynamics
BB3 - structures
$\boldsymbol{\Phi}$ - maximum range for given mission characteristics $\mathrm{Y}_{1,2}$ - includes the aerodynamic drag; $\mathrm{Y}_{1,3}$ - includes the structural weight; $Y_{2,1}$ - includes Mach number; $\mathrm{Y}_{3,1}$ - includes TOGW; $\mathrm{Y}_{2,3}$ - includes the structural deformations that alter the aerodynamic shape; $\mathrm{Y}_{3,2}$ - includes the aerodynamic loads $\mathrm{g}_{1.1}$ - a noise abatement constraint on the mission pro-
file; $g_{2 . t}$ - limit of the chordwise pressure gradient; $\mathrm{g}_{3, \mathrm{t}}$ - allowable stress
$\mathrm{x}_{1, \mathrm{j}}$ - cruise altitude; $\mathrm{x}_{2 \mathrm{j}, \mathrm{j}}$ - leading edge radius; $\mathrm{x}_{3, \mathrm{j}}-$ sheet metal thickness in the wing skin panel No. 138
$z_{1}$ - wing sweep angle; $z_{2}$ - wing aspect ratio; $z_{3}$ - wing airfoil maximum depth-to-chord ratio; $z_{4}$ - location of the engine on the wing

The system in Figure 1 is characterized by BB level design variables X , and by system-level design variables $Z$. As a reference, if an all-in-one optimization were performed, observing the system at a single level and making no distinction between the treatment of X variables and the treatment of $Z$ variables, the problem could be stated

Given: $\quad \mathrm{X}$ and Z
Find: $\quad \Delta X$ and $\Delta Z$

Minimize: $\quad \Phi(\mathrm{X}, \mathrm{Z}, \mathrm{Y}(\mathrm{X}, \mathrm{Z}))$
Satisfy: $\quad G(X, Z, Y(X, Z)$
Since BLISS approaches this optimization by means of a system decomposition, the algorithm depends on the availability of the derivatives of output with respect to input for each BB. That assumes the differentiability of the BB internal relationships to at least the first order. It is immaterial how the derivatives are computed, finite differencing may always be used, but it is expected that in most cases one will utilize one of the more efficient analytical techniques (Adelman and Haftka, 1993).

The algorithm comprises the system analysis and sensitivity analysis, local optimizations inside of the BBs (that includes the BB-internal analyses), and the system optimization. We will not elaborate on SA beyond pointing out that it is highly problem-dependent, and likely to be iterative if there are any non-linearities in the BB analyses. Each pass through the BLISS procedure improves the design in two steps: first by concurrent optimizations of the BBs using the design variables X and holding Z constant; and next, by means of a system-level optimization that utilizes variables $\mathbf{Z}$. We begin with the BB-level optimization.

### 2.1. BB-level (discipline or subsystem) optimizations.

The basis of the algorithm is the formulation of an objective function unique for each BB such that mini-
mization of that function in each BB results in the minimization of the system objective function. To introduce that formulation let us begin with the system objective function (SOF). The SOF is computed as a single output item in one of the BBs; without loss of generality we assume that it is $\mathrm{BB}_{1}$ so that

$$
\begin{equation*}
\Phi=y_{1, i} \tag{1}
\end{equation*}
$$

is one of the elements of $Y_{1}$.
Total derivatives of $Y$ w.r.t. $\mathbf{x}_{\mathrm{r}, \mathrm{j}}, \mathrm{D}\left(\mathrm{Y}, \mathrm{x}_{\mathrm{r}, \mathrm{j}}\right)$, are computed according to Sobieszczanski-Sobieski, 1990, by solving a set of simultaneous, algebraic equations known as Global Sensitivity Equations, GSE, (see Appendix, Section 1, for details) for a particular $\mathrm{X}_{\mathrm{r}, \mathrm{j}}$

$$
\begin{equation*}
[\mathrm{A}]\left\{\mathrm{D}\left(\mathrm{Y}, \mathrm{x}_{\mathrm{r}, \mathrm{j}}\right)\right\}=\left\{\mathrm{d}\left(\mathrm{Y}, \mathrm{x}_{\mathrm{r}, \mathrm{j}}\right)\right\} \tag{2}
\end{equation*}
$$

where $A$ is a square matrix, NYxNY, composed of submatrices forming this pattern

$$
\left[\begin{array}{c|c|c}
\mathbf{I} & \mathbf{A}_{1,2} & \mathbf{A}_{1,3}  \tag{3}\\
\hline \mathbf{A}_{2,1} & \mathbf{I} & \mathbf{A}_{2,3} \\
\hline \mathbf{A}_{3,1} & \mathbf{A}_{3,2} & \mathbf{I}
\end{array}\right]
$$

where I stands for identity matrix, $\mathrm{NY}_{\mathrm{r}} \mathrm{xNY}_{\mathrm{r}}$, and $\mathrm{A}_{\mathrm{r}, \mathrm{s}}$ are matrices of the derivatives that capture sensitivity of the $\mathrm{BB}_{\mathrm{r}}$ output to input. For example

$$
\begin{align*}
& A_{2,3}=-\left[d\left(Y_{2}, Y_{3}\right)\right], N Y_{2} \times N Y_{3} \\
& A_{3,2}=-\left[d\left(Y_{3}, Y_{2}\right)\right], N Y_{3} \times N Y_{2} \tag{4}
\end{align*}
$$

One should note that eq. 2 can be efficiently solved for many different $\mathrm{x}_{\mathrm{r}, \mathrm{j}}$ using techniques available for linear equations with many right-hand sides.

Having $D\left(Y, x_{r}\right)$ computed from eq. 2 for all $x_{r, j}$, we can express $\Phi$ as a function of $X$ by the linear part of the Taylor series

$$
\begin{align*}
\Phi=y_{1, i}= & \left(y_{1, \mathrm{i}}\right)_{0}+\mathrm{D}\left(\mathrm{y}_{1, \mathrm{i}}, \mathrm{X}_{1}\right)^{\mathrm{T}} \Delta \mathrm{X}_{1}+ \\
& \mathrm{D}\left(\mathrm{y}_{1, \mathrm{i}}, \mathrm{X}_{2}\right)^{\mathrm{T}} \Delta \mathbf{X}_{2}+\mathrm{D}\left(\mathrm{y}_{1 ., \mathrm{i}}, \mathrm{X}_{3}\right)^{\mathrm{T}} \Delta \mathrm{X}_{3} \tag{5}
\end{align*}
$$

where D-terms are vectors of length $\mathbf{N X}_{\mathrm{r}}$.
We see from eq. 5 that

$$
\begin{align*}
& \Delta \Phi= \mathrm{D}\left(\mathbf{y}_{1, \mathrm{i}}, \mathbf{X}_{1}\right)^{\mathrm{T}} \Delta \mathbf{X}_{1}+\mathrm{D}\left(\mathrm{y}_{1, \mathrm{i},}, \mathbf{X}_{2}\right)^{\mathrm{T}} \Delta \mathbf{X}_{2}+ \\
& \mathrm{D}\left(\mathrm{y}_{1, i,}, \mathbf{X}_{3}\right)^{\mathrm{T}} \Delta \mathbf{X}_{3} \tag{6}
\end{align*}
$$

the three terms showing explicitly the contributions to $\Delta \Phi$ of the local design variables from each of the three BBs.

It is apparent that to minimize $\Delta \Phi$ we need to charge each BB with the task of minimizing its own objective. Using $\mathrm{BB}_{2}$ as an example, objective $\phi_{2}$ is

$$
\begin{equation*}
\phi_{2}=\mathrm{D}\left(\mathrm{y}_{1, \mathrm{i}}, \mathrm{X}_{2}\right)^{\mathrm{T}} \Delta \mathrm{X}_{2, \mathrm{j}}, \mathrm{j}=1 \cdots \mathrm{NX}_{2} \tag{7}
\end{equation*}
$$

The above equations state mathematically the fundamentally important concept that in a system optimization the contributing disciplines should not optimize themselves for a traditional, disciplinespecific objective such as the minimum aerodynamic drag or minimum structural weight. They should optimize themselves for a "synthetic" objective function that measures the influence of the $\mathrm{BB}_{\mathrm{r}}$ design variables $X_{r}$ on the entire system objective function.

Another way to look at it is to observe that, in longhand

$$
\begin{align*}
\phi_{2} & =\mathrm{D}\left(\mathrm{y}_{1, \mathrm{i}}, \mathrm{x}_{2,1}\right)^{\mathrm{T}} \Delta \mathrm{X}_{2,1}+\mathrm{D}\left(\mathrm{y}_{1, \mathrm{i}}, \mathrm{x}_{2,2}\right)^{\mathrm{T}} \Delta \mathrm{X}_{2,2}+\ldots \\
& +\mathrm{D}\left(\mathrm{y}_{1, \mathrm{j}}, \mathrm{x}_{2, \mathrm{j}}\right)^{\mathrm{T}} \Delta \mathrm{X}_{2, \mathrm{j}}+\ldots, \mathrm{j}=1-->\mathrm{NX}_{2} \tag{8}
\end{align*}
$$

so it may be regarded as a composite objective function commonly used in multiobjective optimization. One may say, therefore, that in a coupled system the local disciplinary or subsystem optimizations should be multiobjective with a composite objective function. The composite objective should be a sum of the local design variables weighted by their influence on the single objective of the whole system. It should be emphasized that this is true also in that particular $\mathrm{BB}_{\mathrm{r}}$ where $\boldsymbol{\Phi}$ is being computed. In the aircraft example it is $\Phi=y_{1, i}$ in $\mathrm{BB}_{1}$ according to eq. 1. However, the $\mathrm{BB}_{1}$ optimization objective is not $\phi_{1}=y_{1, i}$. Instead, it is $\phi_{1}$ from an equation analogous to eq.8.

The local optimization problem may be stated formally for $\mathrm{BB}_{2}$

Given: $\quad X_{2}, Z$, and $Y_{2,1}, Y_{2,3}$
Find: $\quad \Delta \mathbf{X}_{2}$; length $\mathrm{NX}_{2}$
Minimize: $\quad \phi_{2}=\mathrm{D}\left(\mathrm{y}_{1, \mathrm{i}, \mathrm{i}}, \mathrm{X}_{2}\right)^{\mathrm{T}} \Delta \mathrm{X}_{2}$
Satisfy: $\quad \mathrm{G}_{2} \leq 0$, including side-constraints
Incidentally, we adhere to the convention which calls for minimization of the objective function. If the appli-
cation requires that function be maximized, as it does in the example of aircraft range, we convert the objective, e.g., $\Phi=$ - (range).

The optimization problem for $\mathrm{BB}_{1}$, and $\mathrm{BB}_{3}$ are analogous. All three problems being independent of each other may be solved concurrently. This is an opportunity for concurrent engineering and parallel processing.

By solving eq. 9 for all three BBs, we have improvedthe system because, according to eq. 5 and 6 , we have reduced $\Phi$ by $\Delta \Phi$, while satisfying constraints in each BB.

### 2.2. System-level optimization.

So far we have improved the system by manipulating $X$ in the presence of a constant Z . We can score further improvement by exploiting Zs as variables. To do so we need to know how $Z$ influences $\Phi=y_{1, i}$. That is, we need $D\left(y_{1, i}, Z\right)$.

At this point, the BLISS algorithm forks into two alternatives, termed BLISS/A and BLISS/B.

### 2.2.1. BLISS/A

This version of BLISS computes the derivatives of $Y$ with respect to $Z$ by modified GSE, eq.(2.1/2) (equations from other sections are cited in (), the section number given before the $/$ ). The GSE modification accounts for the fact that optimization of a BB turns its $X$ into a function of $Y$ and $Z$ that enter that particular $B B$ as parameters. The modification leads to a new generalization of GSE that takes the following form

$$
\left[M \left\lvert\,\left\{\begin{array}{l}
\mathrm{D}\left(\mathrm{Y}, \mathrm{z}_{\mathrm{k}}\right)  \tag{1}\\
\mathrm{D}\left(\mathrm{X}, \mathrm{z}_{\mathrm{k}}\right)
\end{array}\right\}=\left\{\begin{array}{l}
\mathrm{d}\left(\mathrm{Y}, \mathrm{z}_{\mathrm{k}}\right) \\
\mathrm{d}\left(\mathrm{X}, \mathrm{z}_{\mathrm{k}}\right)
\end{array}\right\}\right.\right.
$$

termed GSE/OS for GSE/Optimized Subsystems. The GSE/OS yields a vector $\mathrm{D}\left(\mathrm{Y}, \mathrm{z}_{\mathrm{k}}\right)$ and $\mathrm{D}\left(\mathrm{X}, \mathrm{z}_{\mathrm{k}}\right)$, and because $\Phi$ is one of the elements of $Y, \Phi=y_{1, i}$, we get the desired derivative $\mathrm{D}\left(\Phi, \mathbf{z}_{\mathrm{k}}\right)$. Derivation and details of the GSE/OS structure, including the definition of the matrix M, are in Section 2 of the Appendix. At this point it will suffice to say that the matrix of coefficients in GSE/OS is populated with $d\left(Y_{r}, Y_{s}\right)$, $d\left(Y_{r}, X_{r}\right)$, and $d\left(X_{r}, Y_{s}\right)$. These terms and the RHS terms of $d\left(Y, z_{k}\right)$ and $d\left(X, z_{k}\right)$ are obtained from the following sources

```
-d(Y}\mp@subsup{Y}{r}{},\mp@subsup{Y}{5}{}),d(\mp@subsup{Y}{r}{},\mp@subsup{X}{r}{}),\mathrm{ and d(Y, (zk) ----- from BBSA
-d(X
```

The terms $d\left(X, z_{k}\right)$ and $d\left(X_{r}, Y_{s}\right)$ are the derivatives of optimum with respect to parameters that, in principle, may be obtained by differentiation of the Kuhn-Tucker conditions, e.g., an algorithm described in Sobieszc-zanski-Sobieski et al, 1982. That approach, however, requires second order derivatives of behavior, too costly in most large-scale applications. Therefore, an approximate algorithm adapted from Vanderplaats and Cai, 1986, is given in Section 3 of the Appendix. In that algorithm, parameters are perturbed by a small increment, one at a time, and the BB optimization is repeated by Linear Programming (LP) starting from the optimal point. Derivatives of optimal $X$ and $Y$ with respect to parameters are then computed by finite differences.

### 2.2.2. BLISS/B

This version of BLISS avoids calculation of $d\left(X_{r}, z_{k}\right)$ and $d\left(X_{r}, Y_{s}\right)$ altogether by using an algorithm that yields $D(\Phi, P)$, where $P$ includes both $Y$ and $Z$. The algorithm, described in literature (e.g., Barthelemy and Sobieszczanski-Sobieski, 1983) is based on the wellknown notion that the Lagrange multipliers may be interpreted as the prices, stated in the units of $\Phi$, for the constraint changes caused by incrementing $\mathrm{p}_{\mathrm{i}}$. For a general case of the objective $F=F(P)$ and $G_{0}=G_{0}(P)$, the algorithm gives the following formula for $\mathrm{D}(\mathrm{F}, \mathrm{P})_{\text {o }}$

$$
\mathrm{D}(\mathrm{~F}, \mathrm{P})_{0}=\mathrm{d}(\mathrm{~F}, \mathrm{P})+\mathrm{L}^{\mathrm{T}} \mathrm{~d}\left(\mathrm{G}_{\mathrm{o}}, \mathrm{P}\right)
$$

To use the above in BLISS, consider that in P we have an independent $Z$ but $Y=Y(Z)$ so that the terms $d()$ require chain-differentiation. Hence, the above general formula tranforms to

$$
\begin{align*}
& \mathrm{D}\left(\mathrm{y}_{1, \mathrm{i}, \mathrm{Z}} \mathrm{Z}_{\mathrm{o}}{ }^{\mathrm{T}}=\left(\mathrm{L}^{\mathrm{T}} \mathrm{~d}\left(\mathrm{G}_{\mathrm{o}}, \mathrm{Z}\right)\right)_{1}+\left(\mathrm{L}^{\mathrm{T}} \mathrm{~d}\left(\mathrm{G}_{\mathrm{o}}, \mathrm{Z}\right)\right)_{2}+\right. \\
& \quad\left(\mathrm{L}^{\mathrm{T}} \mathrm{~d}\left(\mathrm{G}_{\mathrm{o}}, \mathrm{Z}\right)\right)_{3}+\left[\left(\mathrm{L}^{\mathrm{T}} \mathrm{~d}\left(\mathrm{G}_{\mathrm{o}}, \mathrm{Y}\right)\right)_{1}+\left(\mathrm{L}^{\mathrm{T}} \mathrm{~d}\left(\mathrm{G}_{\mathrm{o}}, \mathrm{Y}\right)\right)_{2}\right. \\
& \left.\quad+\left(\mathrm{L}^{\mathrm{T}} \mathrm{~d}\left(\mathrm{G}_{\mathrm{o}}, \mathrm{Y}\right)\right)_{3}\right](\mathrm{D}(\mathrm{Y}, \mathrm{Z}))+\mathrm{D}\left(\mathrm{y}_{1, \mathrm{i}}, \mathrm{Z}\right)^{\mathrm{T}} \tag{1}
\end{align*}
$$

where L is the vector of Lagrange multipliers and ()$_{1}$, ()$_{2}$, and ()$_{3}$ identify the BBs 1,2 , and 3 .

The terms in the above equation originate from the following sources:

- $d\left(G_{0}, Z\right)$ and $d\left(G_{0}, Y\right)$ - BBSA performed on isolated $\mathrm{BB}_{\mathrm{r}}$
- L - obtained for $\mathrm{BB}_{\mathrm{r}}$ at the end of BBOPT
- D(Y,Z) - from GSE in SSA
- $D\left(y_{1, i}, Z\right)$ - the column corresponding to $y_{1, i}$ in the
above matrix $\mathrm{D}(\mathrm{Y}, \mathrm{Z})$

BLISS/B is substantially simpler in implementation than BLISS/A and it eliminates the computational cost of one LP per parameter Y and Z. Optimizers that yield $L$ as a by-product of optimization are available for use in BBOPT, or $L$ may be obtained as described in Haftka and Gurdal, 1992.

### 2.2.3. Optimization in the Z-space.

Once $D\left(y_{1, i}, Z\right)$ have been computed from either eq.(2.2.1/1) or as $D\left(y_{1, i}, Z\right)_{0}$ from eq. (2.2.2/1), we can further improve the system objective by executing the following optimization, using any suitable optimizer

Given: $\quad \mathrm{Z}$ and $\boldsymbol{\Phi}_{0}$
Find: $\quad \boldsymbol{\Delta Z}$
Minimize: $\quad \Phi=\Phi_{0}+\mathrm{D}\left(\mathrm{y}_{1, \mathrm{i}}, \mathrm{Z}\right)^{\mathrm{T}} \Delta \mathrm{Z}$
Satisfy: $\quad \mathrm{ZL} \leq \mathrm{Z}+\Delta \mathrm{Z} \leq \mathrm{ZU} ; \Delta \mathrm{ZL} \leq \Delta \mathrm{Z} \leq \Delta \mathrm{ZU}$
Where $\Phi_{0}$ is inherited from the previous cycle SA for $X$ and $Z$ (initialized if it is the first cycle). It is recommended to handle the $Z$ constraints by means of a trust-region technique, e.g., Alexandrov 1996. In the above, term $\mathrm{D}\left(\mathrm{y}_{1, i}, \mathrm{Z}\right)$ is a constrained derivative that protects $\mathrm{G}_{\mathrm{o}}=0$ in all BBs. Therefore, the optimization is unconstrained except of the side-constraints and move limits.

However, some BBs may have constraints that depend on $Z$ and $Y$ more strongly than on $X$ (in the extreme case some constraints may not be functions of $X$ at all, only of $Y$ and $Z$ ). Such constraints, denoted $G_{y z}$, may be difficult (or impossible) to satisfy by manipulating only X in BBOPT. To satisfy them, one must add to the Z-space optimization in eq. 1 their extrapolated values

$$
\begin{align*}
G_{y z}^{T}= & G_{y z, 0}{ }^{T}+\left(d\left(G_{y z}, Z\right)+\right. \\
& \left.d\left(G_{y z}, Y\right) D(Y, Z)\right)^{T} \Delta Z \leq 0 \tag{2}
\end{align*}
$$

where $d\left(G_{y z}, Z\right)$, and $d\left(G_{y z}, Y\right)$ are obtained from BBSA. In this instance, the $Z$-space optimization becomes a constrained one.

## 3. Iterative Procedure

The two operations, the local optimizations in the BBs and the system-level optimization, described in Sec. 2, result in a new system, altered because of the increments on $X$ and $Z$. This means that inputs to and outputs from SA, BBA, BBSA, SSA, BBOPT, BBOSA (BLISS/A), and SOPT all need to be updated, and the
sequence of these operations repeated with the new values of all quantities involved, including new values of all the derivatives because they would change if there were any nonlinearities in the system (as there usually are).

In a large-scale application where execution of each BLISS cycle may require significant resources and time, the engineering team may wish to review the results before committing to the next cycle. That intervention may entail a problem reformulation, such as overriding the variable values, deleting and adding variables, constraints, and even BBs.


Figure 2: BLISS/B Flowchart
Thus, the following procedure emerges, illustrated also by a flowchart in Figure 2 for BLISS/B with the BLISS/A operations, if different, noted in [ ].
0. Initialize $\mathbf{X}$ \& $\mathbf{Z}$.

1. SA to get Ys and Gs; this includes BBAs for all BBs.
2. Examine TERMINATION CRITERIA, exercise judgment to override the results, modify the problem formulation, and CONTINUE or STOP.
3. BBSA to obtain $d(Y, X), d\left(Y_{r, s}, Y_{s}\right), d(G, Z)$, and $\mathrm{d}(\mathrm{G}, \mathrm{Y})$, and $S S A$, eq. (2.1/2), to get $\mathrm{D}(\mathrm{Y}, \mathrm{X})$ [and
$\mathrm{D}(\mathrm{Y}, \mathrm{Z})]$; Here is an opportunity for concurrent processing.
4. BBOPT for all BBs, eq. (2.1/9) using $\phi$ formulatedindividually for each BB (eq. (2.1/6, 7)), get $\phi_{\text {opt }}$ and $\Delta X_{\text {opt }}$; obtain $L$ for $G_{o}$ [skip L]. Here is an opportunity for concurrent processing.
5. Obtain $\mathrm{D}(\Phi, \mathrm{Z})$ as in eq. (2.2.2/1). [Execute BBOSA to obtain $d(X, Z)$ and $d(X, Y)$, and form and solve GSE/OS (Appendix, Section 3) to generate $D(Y, Z)]$. Here is an opportunity for concurrent processing.
6. SOPT to get $\Delta Z_{\text {opt }}$ by eq. (2.2.3/1 and 2 ) herein.
7. Update all quantities, and repeat from 1.

$$
\mathrm{X}=\mathrm{X}_{0}+\Delta \mathrm{X}_{\mathrm{op}} ; \mathrm{Z}=\mathrm{Z}+\Delta \mathrm{Z}_{\mathrm{opt}}
$$

Note: Termination is placed as \#2 after SA to ensure that the full analysis results document the final system design, as opposed to having it documented only by the extrapolated quantities. Also, at this point the engineering team may decide whether to intervene by modifying the variable values, and adding or deleting the design variables and constraints.

When started from a feasible design, the procedure will result in an improved system, while the local constraints are kept satisfied within extrapolation accuracy, even when terminated before convergence.


Figure 3: Polynomial Representation of Wing Twist
In case of an infeasible design start, the improvement will be in the sense of reductions in the constraint violations, while the objective may exhibit an increase, at least initially. The procedure achieves the improvement by virtue of optimization alternating between the do-
main of NB X-spaces (Step \#4) and the single Z-space (Step \#6).

Caveat: because in BLISS/B the extrapolation of $\Phi$ in eq. (2.2.3/1) is based on the Lagrange multipliers in eq. (2.2.2/1), its accuracy depends on the BBOPT yielding a feasible solution, and on the active constraints $G_{o}$ remaining active for updated $Z$. If some constraints leave the active set $G_{0}$, or new constraints enter, a discontinuous change of the extrapolation error may result. For example, consider the wing aspect ratio AR as a Z-variable and suppose that for $\mathrm{AR}=3$ it is the stress due to the wing bending that is one of the active constraints in the structures BB . If optimization in the Z -space took the design to $\mathrm{AR}=4$, the next cycle may reveal that the stress constraint is satisfied but a flutter constraint becomes critical. Past experience (Sobieszczanski-Sobieski, 1983) shows that this discontinuity is likely to slow, but not to prevent, the process convergence, and may be controlled by adjusting the move limits.


Figure 4: Data Dependencies for Range Optimization

## 4. Numerical Tests and Examples

BLISS/A was tested on a sample of test problems from Hock and Schittkowski, 1981, and on a design of an electronic package. BLISS/B was exercised on the latter, and also on a very simplified aircraft configuration problem. Both versions of BLISS performed as intended in all of the tests. The sole purpose of these initial numerical experiments was to test and to demonstrate the BLISS procedure logic and data flow, therefore, the BBs were merely surrogates of the numerical processes that need to be used in real applications.

### 4.1. Aircraft Optimization

The aircraft test was an optimum cruise segment of a supersonic business jet based on the 1995-96 AIAA Student Competition. This problem was selected because of its available data base and the availability of the black boxes written in Visual Basic in form of Excel spreadsheets. The supersonic business jet was modeled as a coupled system of structures ( $\mathrm{BB}_{1}$ ), aerodynamics ( $\mathrm{BB}_{2}$ ), propulsion ( $\mathrm{BB}_{3}$ ), and aircraft range $\left(\mathrm{BB}_{4}\right)$. All the disciplines were represented by modules comprising an analysis level typical for an early conceptual design stage.

| var \eycle* | 1 | 2 | 3 | 4 | 5 |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Aange (SSA) | 535.79 | 1581.67 | 3425.35 | 3961.41 | 3963.98 |
| Extpl. Error | -535.79 | -536.67 | -431.63 | -56.26 | -3.43 |
| BB1 Extpl. | 17.17 | -0.16 | -3.26 | -0.86 | 0.00 |
| BE2 Exipl. | 16.85 | 0.00 | 0.00 | 0.00 | 0.00 |
| B83 Extpl. | 26.00 | 110.92 | -76.84 | 0.00 | 0.00 |
| X Extpl. | 60.02 | 110.75 | -80.10 | -0.86 | 0.00 |
| $z$ Extpl. | 449.19 | 1301.30 | 559.90 | 0.00 | 0.00 |
| Range (Extpl.) | 1045.00 | 2993.72 | 3905.15 | 3960.55 | 3963.98 |
| $\lambda$ | 0.25 | 0.14951 | 0.17476 | 0.25775 | 0.38757 |
| $x$ | 1 | 0.75 | 0.75 | 0.75 | 0.75 |
| C, | 1 | 0.75 | 0.75 | 0.75 | 0.75 |
| T | 0.5 | 0.1676 | 0.20703 | 0.15624 | 0.15624 |
| t/c | 0.05 | 0.06 | 0.06 | 0.06 | 0.06 |
| h(ft) | 45000 | 54000 | 60000 | 60000 | 60000 |
| M | 1.6 | 1.4 | 1.4 | 1.4 | 1.4 |
| AR | 5.5 | 4.4 | 3.3 | 2.5 | 2.5 |
| $\wedge\left({ }^{\circ}\right)$ | 55 | 66 | 70 | 70 | 70 |
| $\mathbf{S}_{\text {rat }}\left(\mathrm{ft}^{2}\right)$ | 1000 | 1200 | 1400 | 1500 | 1500 |

Table 1: A/C Results for 20\% Move Limit
The aircraft optimization was a maximization of therange computed through the Breguet range equation. For testing purposes, additional design and state variables were introduced in BBs 1 through 3, and functional relationships not present in the original BBs were supplied to reflect what is commonly known about the typical functions involved in design. For example, stress is expected to fall as a reciprocal of the increase of the skin thickness in a wing box. Such relationships were represented by polynomial functions. One plot of such a function is shown in Figure 3, portraying the wing twist as a function of the wing box cross-sectional dimensions scale factor and the wing lift.

Section 4 of the Appendix defines the BBs in this ex ample by their input and output variables, and by the functions that link output to input. Table A1 also identifies local constraints and side constraints. Note that $\mathrm{BB}_{2}$ contains a constraint that does not depend on its X or Y input, thus the Z -space optimization is a constrained one, per eq. (2.2.3/1 \& 2). Side constraints on

Z were judiciously selected to guard against conditions not accounted for in the BBAs. For example, the lower bound of 2.5 on aspect ratio stemmed from the subsonic performance considerations.

| num \den | $\lambda$ | x | C. | T | tc |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $W_{\text {r }}$ | 0.01146 | 1.71536 | 0.01981 | -0.15744 | 0.12714 |
| $W_{F}$ | 0 | 0 | 0 | 0 | 0.72626 |
| $\boldsymbol{\Theta}$ | -0.03342 | 0.19971 | 3.31E-15 | -1.73E-14 | -2.10E-14 |
| L | 0.01146 | 1.71536 | 0.01981 | -0.15744 | 0.12714 |
| D | -4.19E-05 | 0.00581 | 0.12457 | -0.00049 | 0.68108 |
| LD | 0.0115 | 1.7095 | -0.1046 | -0.15694 | -0.54935 |
| SFC | 1.98E-20 | -5.07E-18 | -2.70E-17 | 0.08544 | 0 |
| $\mathrm{W}_{\text {E }}$ | -4.40E-05 | 0.0061 | 0.13083 | -1.03986 | 0.71531 |
| ESF | -4.19E-05 | 0.00581 | 0.12457 | -0.99059 | 0.68108 |
| R | -0.00077 | -0.12692 | -0.12581 | -0.07299 | 0.10115 |
| num \den | h | M | AR | Sweep | $\mathrm{S}_{\text {ram }}$ |
| $W_{T}$ | -0.33831 | 0.31958 | 0.08208 | 0.2537 | 0.55182 |
| $\mathrm{W}_{\text {F }}$ | 0 | 0 | -0.36043 | 0. | 1.09211 |
| $\bigcirc$ | -1.93E-13 | -6.15E-14 | -0.10766 | 3.77E-14 | -0.10766 |
| L | -0.33931 | 0.31958 | 0.08208 | 0.2537 | 0.55182 |
| D | -2.1339 | 2.00984 | 3.37E-06 | -0.83983 | 0.99838 |
| LD | 1.84108 | -1.6507 | 0.08207 | 1.10064 | -0.43675 |
| SFC | 0.12946 | 0.05555 | 2.31E-17 | -1.86E-16 | 0 |
| $\mathrm{W}_{\mathrm{E}}$ | -2.24115 | 2.11086 | 3.54E-06 | -0.88204 | 1.04857 |
| ESF | -2.1339 | 2.00984 | 3.37E-06 | -0.83983 | 0.99838 |
| R | 2.07616 | -1.04784 | -0.39618 | 0.82904 | 0.15535 |

Table 2: Normalized Y Derivatives w.r.t. X and Z
The BBs are coupled by the output-to-input data transfers (design structure matrix) depicted in Figure 4. Note that $\mathrm{BB}_{4}$ is an analysis-only module and does not feedback any data to other BBs.


Figure 5: Range and Extrapolation Error Histogram
This test was conducted entirely using MATLAB 5 and its Optimization Toolbox. The entire MATLAB code listing for the aircraft range model may be found in Section 5 of the Appendix. To establish a benchmark, the system was first optimized using an all-in-one approach in which the MATLAB optimizer was coupled directly to SA and saw no distinction between the $X$ and $Z$ variables. Next, the test case was executed using

BLISS/B, starting at different infeasible initial points chosen by varying the six design variables that are not arguments in the polynomial functions. The choice of initial values for variables that are arguments of the polynomial functions was limited due to the nature of the polynomial formulation. This limitation is not a characteristic of the BLISS method itself, as the polynomial functions would not be required in a large scale optimization problem. With the move limits ranging from 10 to $70 \%$, the procedure convergence was satisfactory through the move limits of $60 \%$ for all initial points tested. However, in nearly all cases, no additional improvement in convergence rate was recorded for move limits greater than $20 \%$. For instance, the objective function was advanced to within $1 \%$ of the benchmark in 5 passes for move limits 20 and $30 \%$. Onset of an erratic behavior was observed with move limits increased past $60 \%$, the procedure converged or diverged dependent on the starting point.


Figure 6: Range Sensitivities ( ${ }^{\text {st }}$ cycle)

Table 1 displays a sample of typical results for the move limits value of $20 \%$. It shows that the initial design range was extremely poor, only 536 nm . BLISS/B improvements advanced the range to 3964 nm . The range converged monotonically, although in some cases small amplitude oscillations were observed. Comparison of the extrapolated and actual values of the objective and constraints shows reasonable accuracy and conservatism of the extrapolations. The optimal values of the design variables reflect numerous tradeoffs typical for aircraft design. For instance, opti$\mathrm{mal} \mathrm{t} / \mathrm{c}$ resulted, in part, from a trade-off between the wave drag and structural weight. Table 2 shows normalized (logarithmic) derivatives of all Ys, including the range, w.r.t. all the X and Z variables, sampled in Cycle 1 to illustrate sensitivity of the system solution to design variables.

Figure 5 illustrates the range histogram, and depicts the extrapolation error as being effectively controlled by the move limits. Range sensitivities to $X$ and $Z$ variables are shown in Figure 6. As expected, altitude and Mach number have the largest effect on the objective function, while taper ratio has the smallest.


Figure 7: BB and System Contributions to Range
Figure 7 shows the individual BB and system contributions to the range objective in each cycle. Here it is observed that, in this particular case, the contribution of system level variables is significantly larger than that of the local variables in the extrapolation of range.

This test case was also implemented in a software package for system analysis and optimization called iSIGHT (iSIGHT, 1998). The iSIGHT and MATLAB results cross-check was completely satisfactory.

### 4.2 Electronic Package optimization

The electronic packaging was introduced as an MDO problem in Renaud, 1993. Its electrical and thermal subsystems are coupled because component resistance is influenced by operating temperatures and the temperatures depend on resistance.

The objective of the problem is to maximize the watt density for the electronic package subject to constraints. The constraints require the operation temperatures for the resistors to be below a threshold temperature and the current through the two resistors to be equal. The system diagram in Figure 8 shows the data dependencies for two BBs , representing electrical resistance analysis and thermal analysis. As Figure 8 indicates, there are no "natural" Z's in this case. Therefore, $Z$ 's were created as targets imposed on each
of the Y's and the BBOPT's were required to match the Y values to those Z targets (similar as it is done in the Collaborative Optimization method). Details of the electronic packaging problem may be found in Padula, 1996.


Figure 8: Electronic Packaging Data Dependencies
This test case was implemented in iSIGHT using BLISS/A and B. A benchmark result was obtained by executing an all-in-one optimization from various starting points ("A-in-O" column). BLISS/A and B were started from the same points. Table 4 displays the benchmark and the BLISS/A and B results as showing a good agreement. Table 4 also indicates a comparison of the computational labor (the "Work" column) measured by the number of BB evaluations necessary to converge the fixed-point iterations in BBAs and in SA, all repeated as needed to compute derivatives by finitedifferences in a gradient-guided optimization. As Table 4 shows, the BLISS/B computational labor was substantially lower than the benchmark in all cases.

| Method | Case | Initial Deaign Objective | hit. Des. Max Constr. Viol. | Final Design Objective | Fin. Des. Max Constr. Viol. | Work |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| A-in-O | 1 | 7.79440E+01 | $2.16630 \mathrm{E}-08$ | $6.39720 \mathrm{E}+05$ | 1.22E-03 | 498 |
|  | 2 | 6.83630E+03 | -2.89560E-01 | $6.39720 \mathrm{E}+05$ | 1.22E-03 | 264 |
|  | 3 | $1.51110 \mathrm{E}+03$ | -4.29240E-02 | 6.36540E+05 | 1.45E-03 | 264 |
|  | 4 | $1.46070 \mathrm{E}+01$ | -1.02490E-03 | 6.36940E +05 | $1.42 \mathrm{E}-03$ | 175 |
| BLISS/A | 1 | $7.79440 \mathrm{E}+01$ | $2.16630 \mathrm{E}-08$ | 6.39700E+05 | 1.20E-03 | 436 |
|  | 2 | $6.83630 \mathrm{E}+03$ | -2.89560E-01 | 6.39050E+05 | $1.18 \mathrm{E}-03$ | 508 |
|  | 3 | $1.51110 \mathrm{E}+03$ | -4.29240E-02 | $6.39050 \mathrm{E}+05$ | -4.89E-04 | 174 |
|  | 4 | $1.46070 \mathrm{E}+01$ | -1.02490E-03 | $6.39290 \mathrm{E}+05$ | 3.70E-04 | 313 |
| BLESSB | 1 | $7.79440 \mathrm{E}+01$ | $2.16630 \mathrm{E}-08$ | 6.35720E+05 | 1.22E-03 | 365 |
|  | 2 | $6.83630 \mathrm{E}+03$ | -2.89560E-01 | 6.39720E+05 | $1.22 \mathrm{E}-03$ | 207 |
|  | 3 | $1.51110 \mathrm{E}+03$ | -4.29240E-02 | 6.39720E+05 | 1.22E-03 | 114 |
|  | 4 | $1.46070 \mathrm{E}+01$ | -1.02490E-03 | 6.39720E+05 | 1.22E-03 | 105 |

Table 4: Electronic Packaging Data

## 5. BLISS Status, Assessment, and Concluding Remarks

A method for engineering system optimization was developed to decompose the problem into a set of local optimizations (large number of detailed design variables) and a system-level optimization (small number
of global design variables). Optimum sensitivity data link the subsystem and system level optimizations. There are two variants of the method, BLISS/A and BLISS/B, that differ by the details of that linkage. In the paper, the method algorithm was laid out in detail for a system of three subdomains (modules). Its generalization to NB subdomains is straightforward. The same algorithm may be used to decompose any of the local optimizations, hence optimization may be conducted at more than two levels.

MATLAB and iSIGHT programming languages were used to implement and test the method prototype on a simplified, conceptual level supersonic business jet design, and a detailed design of an electronic device. Dimensionality and complexity of the preliminary test cases was intentionally kept very low for an expeditious assessment of the method potential before more resources are invested in further development. Favorable agreement with the benchmark results and a satisfactory convergence observed in the above tests provided motivation for such development and future testing in larger applications.

Assessment of BLISS at the above development status is as follows. BLISS relies on linearization of a generally nonlinear optimization, therefore its effectiveness depends on the degree of nonlinearity. As any gradi-ent-guided method, it guarantees a cycle-to-cycle improvement, but if the problem is non-convex, its convergence to the global optimum depends on the starting point and may strongly depend on the move limits. In this regard, BLISS's strong points are in the procedure being open to human intervention between the cycles and in the autonomy of the subdomain optimizations in local variables. These optimizations may be conducted by any means deemed to be most suitable by disciplinary experts, hence non-convexity, and strong nonlinearities in terms of the local variables often encountered in subdomains, e.g., the local buckling in thin-walled structures, are isolated and prevented from slowing down the system-level optimization convergence. On the other hand, the optimization robustness may be adversely affected by the local constraints leaving and entering the active constraint set. Effect of the above on BLISS/A is much less than on BLISS/B. This is probably the only reason to continue the development of BLISS/A alongside with BLISS/B, even though BLISS/B has a distinct advantage in simplicity and a much lower computational cost. Once there is more information on the relative merits and demerits of both variants, the better variant may be selected.

The demand BLISS puts on the computer storage is the
same the subdomains would require for their own, stand-alone optimizations, with exception of the generation and solution of the Global Sensitivity Equations. If there is a pair of BBs that exchange large number of the $\mathrm{y}_{\mathrm{r}, \mathrm{s}, \mathrm{i}}$ - quantities, dimensionality of the corresponding matrices that store the derivatives, and computational cost of these derivatives needed to form GSE, may become prohibitive. Some relief may be provided here by application of condensation techniques and by deleting from GSE those derivative matrices that are known to have negligible effect on the system behavior.

The principal advantage of BLISS appears to lie in its separating overall system design considerations from the considerations of the detail. This makes the resulting mapping of its algorithm fit well on diverse, and potentially dispersed, human organizations. This advantage remains to be demonstrated in further development toward large-scale, complex applications.

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## Appendix

This Appendix provides details of the Global Sensitivity Equations (GSE) applied to a system which optimizes BBs, the details of a technique for the BB Optimum Sensitivity Analysis, and the details of the aircraft range optimization model.

## 1. Global Sensitivity Equations

Derivatives of Y w.r.t. X, and Z, are obtained rigorously from the Implicit Function Theorem in Sobieszc-zanski-Sobieski,1990. The condensed derivation is provided below. It begins by recognizing that

$$
\begin{align*}
& Y_{1,2}=Y_{1,2}\left(Z, X_{2,}, Y_{2,1}, Y_{2,3}\right)  \tag{A1}\\
& Y_{1,3}=Y_{1,3}\left(Z, X_{3}, Y_{3,1}, Y_{3,2}\right) \\
& Y_{2,1}=Y_{2,1}\left(Z, X_{1}, Y_{1,2}, Y_{1,3}\right)  \tag{A3}\\
& Y_{2,3}=Y_{2,3}\left(Z, X_{3}, Y_{3,1}, Y_{3,2}\right)  \tag{A4}\\
& Y_{3,1}=Y_{3,1}\left(Z, X_{1}, Y_{1,2}, Y_{1,3}\right)  \tag{A5}\\
& Y_{3,2}=Y_{3,2}\left(Z, X_{2}, Y_{2,1}, Y_{2,3}\right) \tag{A6}
\end{align*}
$$

where the independent variables are X and Z .
Observe that eq. A1-A6 are coupled by $Y_{i .}$ e.g., $Y_{3,1}$ depends on $Y_{1,3}$ in eq. A5, and $Y_{1,3}$ depends on $Y_{3,1}$ in eq. A2. Consider for an example, the chaindifferentiation w.r.t. $\mathrm{x}_{1, \mathrm{j}}$ applied to eq. A3. It yields

A7)

$$
\begin{aligned}
& \mathrm{D}\left(\mathrm{Y}_{2,1}, \mathrm{x}_{1, \mathrm{j}}\right)=\mathrm{d}\left(\mathrm{Y}_{2,1}, \mathrm{x}_{1, j}\right)+ \\
& \mathrm{d}\left(\mathrm{Y}_{2,1}, \mathrm{Y}_{1}\right) \mathrm{D}\left(\mathrm{Y}_{1}, \mathbf{x}_{1, \mathrm{j}}\right)+ \\
& \quad \mathrm{d}\left(\mathrm{Y}_{2,1}, Y_{3}\right) \mathrm{D}\left(\mathrm{Y}_{3}, \mathbf{x}_{1, \mathrm{j}}\right)
\end{aligned}
$$

Repeating the above for the remaining equations, treating $Y_{2,1}$ as a subset of $Y_{1}$, and collecting the terms leads to eq. (2.1/2 and 3 ).

The derivatives of $Y$ w.r.t. $z_{k}$ are obtained by simply replacing $x_{r, j}$ with $z_{k}$ in eq. (2.1/2) to obtain

A8) $\quad[\mathrm{A}]\left\{\mathrm{D}\left(\mathrm{Y}, \mathrm{z}_{\mathrm{k}}\right)\right\}=\left\{\mathrm{d}\left(\mathrm{Y}, \mathrm{z}_{\mathrm{k}}\right)\right\}$

## 2. GSE/Optimized Subsystems

In the preceding section both X and Z are independent variables. By virtue of BBOPT conducted for constant $Z$ and $Y$ inputs, $X$ becomes dependent on $Z$ so that derivatives of $X$ w.r.t. exist in addition to derivatives of Y w.r.t. Z. For example, optimal $X_{2}$ depends on $Z$, $Y_{2,1}$, and $Y_{2,3}$, that are parameters in the optimization of $\mathrm{BB}_{2}$. Hence, to compute the derivatives of Y and X w.r.t. $Z$, we begin by rewriting the functional relationships in eq. A1-A6, adding the new dependencies in all three BBs in the system,

$$
\begin{equation*}
Y_{1,2}=Y_{1,2}\left(Z, X_{2,}, Y_{2,1}, Y_{2,3}\right) \tag{A9}
\end{equation*}
$$

(A10)

The same Implicit Function Theorem that is the basis of the GSE derivation may be applied to the above equations to obtain $D(Y, Z)$. For example, by applying chain-differentiation to $\mathrm{Y}_{2,1}$ treated as a subset of $\mathrm{Y}_{2}$, we obtain
(A18)

$$
\begin{gathered}
\mathrm{D}\left(\mathrm{Y}_{2}, \mathrm{Z}_{\mathrm{k}}\right)=\mathrm{d}\left(\mathrm{Y}_{2}, \mathrm{Z}_{\mathrm{k}}\right)+\mathrm{d}\left(\mathrm{Y}_{2}, \mathrm{X}_{2}\right) \mathrm{D}\left(\mathrm{X}_{2}, \mathrm{Z}_{\mathrm{k}}\right)+ \\
\mathrm{d}\left(\mathrm{Y}_{2}, \mathrm{Y}_{1}\right) \mathrm{D}\left(\mathrm{Y}_{1}, \mathrm{Z}_{\mathrm{k}}\right)+ \\
\mathrm{d}\left(\mathrm{Y}_{2}, \mathrm{Y}_{3}\right) \mathrm{D}\left(\mathrm{Y}_{3}, \mathrm{Z}_{\mathrm{k}}\right)
\end{gathered}
$$

and for $\mathrm{X}_{2}$, again as one example:

$$
\begin{gather*}
\mathrm{D}\left(\mathrm{X}_{2}, \mathrm{Z}_{\mathrm{k}}\right)=\mathrm{d}\left(\mathrm{X}_{2}, \mathrm{Z}_{\mathrm{k}}\right)+\mathrm{d}\left(\mathrm{X}_{2}, \mathrm{Y}_{1}\right) \mathrm{D}\left(\mathrm{Y}_{1}, \mathrm{z}_{\mathrm{k}}\right)+  \tag{A19}\\
\mathrm{d}\left(\mathrm{X}_{2}, \mathrm{Y}_{3}\right) \mathrm{D}\left(\mathrm{Y}_{3}, \mathrm{z}_{\mathrm{k}}\right)
\end{gather*}
$$

In the above, the D-terms are the total derivatives we seek, while the d-terms are the partial derivatives of two, different kinds. The derivatives of $Y_{r}$ w.r.t. $Y_{S}$ and $Y_{r}$ w.r.t. $X_{r}$ are obtained from BBSA $_{r}$ using any sensitivity analysis algorithm appropriate for the particular $\mathrm{BB}_{\mathrm{r}}$ (including the option of finite differencing). The derivatives of $X_{r}$ w.r.t. $\mathbf{z}_{\mathrm{k}}$ and $X_{r}$ w.r.t. $Y_{s}$ are produced by an analysis of optimum for sensitivity to parameters, BBOSA $_{r}$, explained in later in this Appendix.

As a mathematical digression, one should mention at this point that the derivatives termed partial in the above would be called total in both BBSA and BBOSA. This is not a contradiction. It is so because the partial and total derivatives are hierarchically related in a multilevel system of parents and children. What is a total derivative in a child is partial at the parent level. In the application herein, the system of coupled three BBs is a parent, each BB is a child.

The chain-derivative expressions for $Y_{1}, Y_{3}, X_{1}$ and $X_{3}$ look similar to eq. A18 and A19, differences are only in the subscripts. When the entire set of six chain-
derivative expressions is written it forms a set of simultaneous, algebraic equations in which the total derivatives such as $D\left(Y_{2}, Z_{k}\right)$ and $D\left(X_{2}, Z_{k}\right)$ appear as unknowns. This is a new generalization of GSE, termed GSE/OS for GSE/Optimized Subsystems. For the case of three-BB system, these equations may be presented in a matrix format like this

$$
\begin{aligned}
& \text { (A20) } \\
& {\left[\mathbf{M}_{\mathrm{yy}}\right]\left\{\mathrm{D}\left(\mathrm{Y}, \mathrm{z}_{\mathrm{k}}\right)\right\}+\left[\mathrm{M}_{\mathrm{yx}}\right]\left\{\mathrm{D}\left(\mathrm{X}, \mathrm{z}_{\mathrm{k}}\right)\right\}=\mathrm{d}\left(\mathrm{Y}, \mathrm{z}_{\mathrm{k}}\right)} \\
& {\left[\mathbf{M}_{\mathrm{xy}}\right]\left\{\mathrm{D}\left(\mathrm{Y}, \mathbf{z}_{\mathbf{k}}\right)\right\}+\left[\mathrm{M}_{\mathrm{x} \mathbf{x}}\right]\left\{\mathrm{D}\left(\mathbf{X}, \mathrm{z}_{\mathbf{k}}\right)\right\}=\mathrm{d}\left(\mathbf{X}, \mathbf{z}_{\mathbf{k}}\right)}
\end{aligned}
$$

The internal structure of the M-matrices in the above is
for $\left[\mathrm{M}_{\mathrm{yy}}\right]$ :

$$
\left[\begin{array}{c|c|c}
I & -d\left(Y_{1}, Y_{2}\right) & -d\left(Y_{1}, Y_{3}\right) \\
\hline-d\left(Y_{2}, Y_{1}\right) & I & -d\left(Y_{2}, Y_{3}\right) \\
\hline-d\left(Y_{3}, Y_{1}\right) & -d\left(Y_{3}, Y_{2}\right) & I
\end{array}\right]
$$

for $\left[\mathrm{M}_{\mathrm{yx}}\right]$ :

$$
\left[\begin{array}{c|c|c}
-\mathrm{d}\left(Y_{1}, X_{1}\right) & 0 & 0 \\
\hline 0 & -\mathrm{d}\left(Y_{2}, X_{2}\right) & 0 \\
\hline 0 & 0 & -\mathrm{d}\left(\mathbf{Y}_{3}, X_{3}\right)
\end{array}\right]
$$

for $\left[M_{x y}\right]$ :

$$
\left[\begin{array}{c|c|c}
0 & -\mathrm{d}\left(\mathbf{X}_{1}, Y_{2}\right) & -\mathrm{d}\left(\mathbf{X}_{1}, Y_{3}\right) \\
\hline-\mathrm{d}\left(X_{2}, Y_{1}\right) & 0 & -\mathrm{d}\left(\mathbf{X}_{2}, Y_{3}\right) \\
\hline-\mathrm{d}\left(\mathbf{X}_{3}, Y_{1}\right) & -\mathrm{d}\left(X_{3}, Y_{2}\right) & 0
\end{array}\right]
$$

and for $\left[M_{x x}\right]$ :

$$
\left[\begin{array}{c|c|c}
\mathbf{1} & 0 & 0 \\
\hline 0 & 1 & 0 \\
\hline 0 & 0 & \mathbf{I}
\end{array}\right]
$$

Again, in the above, all $Y_{r, s}$ are folded into $Y_{r}$ for compactness, and the terms are falling into the previously introduced categories as follows:

- $\mathbf{M}_{\mathrm{yy}}, \mathbf{M}_{\mathrm{yx}}$, and $\mathrm{d}\left(\mathrm{Y}, \mathrm{z}_{\mathrm{k}}\right)$------ from BBSA
- $\mathrm{M}_{\mathrm{xy}}$ and $\mathrm{d}\left(\mathrm{X}, \mathrm{z}_{\mathrm{k}}\right)$----------- from BBOSA

As in GSE, one may obtain $\mathrm{D}\left(\mathrm{Y}_{2}, \mathrm{z}_{\mathrm{k}}\right)$ and $\mathrm{D}\left(\mathrm{X}_{2}, \mathrm{z}_{\mathrm{k}}\right)$ for all $z_{k}, k=1 \ldots>N Z$ by means of one of the efficient techniques for linear equations with many right-handsides.

## 3. Black Box Optimum Sensitivity Analysis (BBOSA)

Analysis of optimum for sensitivity to parameters (also called the post-optimum analysis) is preceded by solving a BB optimization problem
(A21) Given: $P$
Find: $\quad \mathbf{X}$
Minimize: (X,P)
Satisfy: $\quad G(X, P) \leq 0$, including sideconstraints and move limits
where $P$ are parameters kept constant while an optimizer manipulates $\mathbf{X}$.

In the BLISS application, the parameters $P$ in $B_{r}$ are
$z_{k}$, and $Y_{r, s}$, because these quantities are kept constant in $\mathrm{BBOPT}_{\mathrm{r}}$.

After $\phi_{\min }$, and $X_{\text {opt }}$ are found, one may seek sensitivity of these quantities to the change of $P$ in form of the derivatives $\mathrm{D}\left(\phi_{\min }, \mathrm{P}\right)$ and $\mathrm{D}\left(\mathrm{X}_{\mathrm{opl}}, \mathrm{P}\right)$.

Vanderplaats and Cai, 1986, review techniques, rigorous and approximate, available for calculating $\mathrm{D}\left(\mathrm{X}_{\mathrm{op} 1}, \mathrm{P}\right)$. The technique adapted for the BLISS/A purposes comprises the following steps executed for $\mathrm{BB}_{\mathrm{r}}$ :

Table A1: BB Definitions

linearly and by Linear Programming solve

| Find | $X$ |
| :--- | :--- |
| Minimize | $F$ |
| Satisfy | $G_{0} \leq 0$ |

$\mathrm{XL}<=\mathrm{X}<=\mathrm{XU}$; where XL and XU incorporate the side constraints and the move limits;
to obtain $\mathrm{X}_{\mathrm{opt}}$.
3. Approximate $\mathrm{D}(\mathrm{X}, \mathrm{P})=\Delta \mathrm{X}_{\text {opl }} / \Delta \mathrm{P}$
4. Repeat from \#1 for all elements of $Z$ and $Y$ input into $\mathrm{BB}_{\mathrm{r}}$.

Repeated for all BBs, the above procedure yields a set of $D(X, Z)$ and $D(X, Y)$ to be entered as $d(X, Z)$ and $\mathrm{d}(\mathrm{X}, \mathrm{Y})$ into GSE/OS, eq. A20. Solution of eq. A20 provides $D(X, Z)$ and $D(Y, Z)$. The latter is substituted into eq. (2.2.3/2), and $D(\Phi, Z)$, extracted from $D(Y, Z)$, goes into eq. (2.2.3/1).

## 4. A/C Range Optimization Model.

Table A1 shows the equations used in each of the BBs for the aircraft model. Polynomial functions are represented by 'pf()' with independent variables in the parentheses. Each polynomial function is of the form:

$$
\begin{equation*}
P F=A_{o}+A_{i} * S^{T}+(1 / 2) * S^{*} A_{i j} * S^{T} \tag{A22}
\end{equation*}
$$

Where $S$ is the vector of independent variables, and $A_{0}$, $\mathrm{A}_{\mathrm{i}}$, and $\mathrm{A}_{\mathrm{ij}}$ are coefficient terms.

In calculating the polynomial functions using eq. A22, terms in the $S$ vectors are in the same order as they appear in pf() in Table A1. The off diagonal terms of $\mathrm{A}_{\mathrm{ij}}$ are random numbers between 0 and 1 . For this model, they are

$$
\mathrm{A}_{\mathrm{ij}}=\left[\begin{array}{c|c|c|c|c}
-- & 0.3970 & 0.8152 & 0.9230 & 0.1108 \\
\hline 0.4252 & -- & 0.6357 & 0.7435 & 0.1138 \\
\hline 0.0329 & 0.8856 & -- & 0.3657 & 0.0019 \\
\hline 0.0878 & 0.7248 & 0.1978 & -- & 0.0169 \\
\hline 0.8955 & 0.4568 & 0.8075 & 0.9239 & --
\end{array}\right]
$$

The remaining coefficient are:
$-\quad \Theta \cdots \quad A_{0}=[1.0] ; A_{i}=[0.3-0.3-0.3-0.2] ;$
$\mathrm{A}_{\mathrm{ij}}=[0.4-0.4-0.40]$;

- Fol $--->\quad A_{0}=[1.0] ; A_{i}=[6.25] ; A_{i i}=[0] ;$
- $\sigma 1-->\quad A_{o}=[1.0] ; A_{i}=[-0.750 .5-0.750 .5$
0.5]; $\mathrm{A}_{\mathrm{ii}}=\left[\begin{array}{lllll}-2.5 & 0 & -2.5 & 0 & 0\end{array}\right] ;$
- $\sigma 2 \cdots \quad A_{0}=[1.0] ; A_{i}=[-0.50 .333-0.50 .333$
$0.333] ; \mathrm{A}_{\mathrm{ij}}=\left[\begin{array}{lll}-1.111 & 0-1.11100 & 0\end{array}\right.$;
- $\sigma 3-->\quad \mathrm{A}_{0}=[1.0] ; \mathrm{A}_{\mathrm{i}}=[-0.3750 .25-0.375$
$0.250 .25] ; \mathrm{A}_{\mathrm{ii}}=\left[\begin{array}{llll}-0.625 & 0-0.625 & 0 & 0\end{array}\right] ;$
- $\sigma 4-->\quad A_{0}=[1.0] ; A_{i}=[-0.30 .2-0.30 .20 .2]$;
$\mathrm{A}_{\mathrm{ii}}=\left[\begin{array}{lll}-0.4 & 0-0.4 & 0\end{array}\right]$;
- $\sigma 5 \cdots \quad \mathrm{~A}_{0}=[1.0] ; \mathrm{A}_{\mathrm{i}}=[-0.250 .1667-0.25$
$0.16670 .1667] ; \mathrm{A}_{\mathrm{ij}}=[-0.27780$
-0.277800 ];
- Fo2 $-->\quad \mathrm{A}_{\mathrm{o}}=[1.0] ; \mathrm{A}_{\mathrm{i}}=[0.20 .2] ; \mathrm{A}_{\mathrm{ij}}=[00] ;$
- Fo3 ---> $\mathrm{A}_{\mathrm{o}}=[1.0] ; \mathrm{A}_{\mathrm{i}}=[0] ; \mathrm{A}_{\mathrm{ii}}=[0.04]$;
- $\mathrm{dp} / \mathrm{dx} \cdots \mathrm{A}_{0}=[1.0] ; \mathrm{A}_{\mathrm{i}}=[0.2] ; \mathrm{A}_{\mathrm{ij}}=[0]$;
- Temp $-->\mathrm{A}_{0}=[1.0] ; \mathrm{A}_{\mathrm{i}}=[0.3-0.30 .3]$;
$\mathrm{A}_{\mathrm{ii}}=\left[\begin{array}{lll}0.4-0.4 & 0.4\end{array}\right]$;
Equations for SFC and the upper constraint bound on throttle setting in the Propulsion BB are polynomials representing surfaces fit to engine deck data (AIAA UTC/Pratt \& Whitney, 1995/96).


## 5. A/C Range MATLAB Code.

Included in the following pages is the MATLAB code for the aircraft range optimization model. The constrained optimization routine used in BB1OPT, BB2OPT, BB3OPT, and SYSOPT may be found in MATLAB's Optimization Toolbox and is based on a Sequential Quadratic Programming method. The finite differencing subfunctions in FIN_DIFF are simple onestep forward finite difference codes that use a 1 percent step increment.
\％them．Local optimizations are performed on each BB（BBOPT）as well as a

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vary
vary
vary
vary $\stackrel{\lambda}{7}$ vary $\stackrel{\rightharpoonup}{7}$高六荀 축方 vary

| $\%$ |  |  |
| :---: | :---: | :---: |
| \％ | Program Listing |  |
| \％ |  |  |
| \％ | Name | Page Number |
| \％ |  |  |
| \％ | BLISS． | ．．．． 16 |
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| \％ | SYSWRAPPER． | ．． 33 |
| \％ | INBOUNDS．．． | $\ldots$ |
| \％ | FIN＿DIFF．．．．． | $\ldots . . .34$ |
| \％ |  |  |
| \％The electronic version of this code has been placed in custody of Dr．Jaroslaw |  |  |
| \％Sobieski，NASA Langley Research Center，Hampton，VA 23681. |  |  |
|  |  |  |  |  |
|  |  |  |



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Dphi1_Z = [Lagrangel ${ }^{*}$ dg $1 \_$Z $]$;
Dphil_Y = [Lagrange 1**g1_YE1];
Dphi2_Y = [Lagrange2**dg2_YE2];
Dphi3_Y = [Lagrange3**dg3_YE3];

| \% | Z(2) - | - Altitude | ft |
| :---: | :---: | :---: | :---: |
| \% | $\mathrm{Z}(3)$ - | - Mach number | none |
| \% | Z(4) - | - Aspect ratio | none |
| \% | Z(5) - | - Wing sweep | deg |
| \% | $\mathrm{Z}(6)$ - | - Wing surface area | $\mathrm{ft}^{2}$ |
| \% |  |  |  |
| \% | Output Variables |  |  |
| \% | C - | - Vector of constants | vary |
| \% | $\mathrm{G} a \quad$ - | - Vector of constraint values in $\mathrm{BB} a(a=1,2,3)$ | vary |
| \% | Ya | Vector of behavior variables output from from $\mathrm{BB} a(a=1,2,3)$ |  |
| \% | Yab | - Vector of behavior variables output from |  |
| \% |  | $\mathrm{BB} a$, input to $\mathrm{BB} b(a \& b=1,2,3)$ | vary |
| \% | Y4 <br> var_inititial | - Objective function output from BB4 | NM |
| \% |  | preserved values for polynomial construction (var differs depending on particular poly.) | vary |
| \% |  |  |  |
| \% | Local Variables |  |  |
| \% | Lu | - Test variable used in G-S iteration for convergence of lift | lb |
| \% | Weu | - Test variable used in G-S iteration for convergence of engine weight | lb |
| \% | ESFu | Test variable used in G-S iteration for convergence of engine scale factor | none |
| \% |  |  |  |
| \% | Subfunctions : | : |  |
| \% | BB_weight | -Calculates a/c structural weights |  |
| \% | BB_dragpolar | ar -Calculates aerodynamic values |  |
| \% | BB_power | -Calculates propulsion values |  |
| \% | BB_range | -Calculates system objective function |  |
| \% |  |  |  |

CLB = 'Pass_1 Pass_2 Pass_3 Pass_4 Pass_5 Pass_6 Pass_7 Pass_8 Pass_9 Pass_10 Pass_11 Pass_12 Pass_13 Pass_14 Pass_15 Pass_16 Pass_17 Pass_18 Pass_19 Pass_20 Pass_21 Pass_22 Pass_23 Pass_24 Pass_25 Pass_26 Pass_ 27 Pass_ 28 Pass_29 Pass_30 Pass_31 Pass_32 Pass_33 Pass_34 Pass_35 Pass_36 Pass_37 Pass_38 Pass_39 Pass_40';

RLB1 = 'Y1(1) Y1(2) Y1(3) Y2(1) Y2(2) Y2(3) Y3(1) Y3(2) Y3(3) Y4'; CLB1 = 'X1(1) X1(2) X2 X3';
\%printmat(DYX_nd(:,:,1), Non-Dimensional D(Y,X)',RLB1,CLB1);
RLB2 = 'Y1(1) Y1(2) Y1(3) Y2(1) Y2(2) Y2(3) Y3(1) Y3(2) Y3(3) Y4';
(YZ), RLB2 CLB2);
$\mathrm{G}=\left[\mathrm{BB} 1 \_\mathrm{G}\right.$ G_sys(:,1) BB3_G];
RLB3 = Pass_1 Pass_2 Pass_3 Pass_4 Pass_5 Pass_6; $\quad$ ESF 1 temp Throttle';
CLB3 = 'sig1 sig2 sig 3 sig4 sig5 twist_u twist_1 dp/dx ESF_u ESF_1 temp Throttle';
printmat(G,'Constraints at Beginning of Pass',RLB3,CLB3);

## Subfunction SYSTEM_ANALYSIS

This subfunction uses Gauss-Seidel iteration on the aircraft range optimization model to compute behavior variables, given a set of design variables. Black boxes WEIGHT, DRAGPOLAR, and POWER are called.
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vary
none
p.f.

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$\stackrel{0}{\circ}$

Wing taper ratio
 - Skin friction coefficient as poly. funct.

Throttle setting

- Thickness/chord ratio
: Jeremy S. Agte


[Y1,Y2,Y3,Y4,Y12,Y14,Y21,Y23,Y24,Y31,Y32,Y34,C]=Y_variables;

| file = 'in_out 1. dat'; write_var(Z,Y1,Y2,Y3,Y4,Y12,Y14,Y21,Y23,Y24,Y31,Y32,Y34,X1,X2,X3,C,file) |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
| \% |  |  |  |  |
| \% |  |  |  |  |
| \% |  | Subfunction BB_WEIGHT |  |  |
| \% |  |  |  |  |
| \% | This subfunction calculates the weight of the aircraft by structure and adds them to obtain a total aircraft weight. It calls the subfunction POLYAPPROX to compute functions represented by polynomials. |  |  |  |
| \% |  |  |  |  |
| \% |  |  |  |  |
| \% | Author | Jeremy S. Agte | NASA Langley/GWU S | Spring 98 |
| \% | Input Variables |  |  |  |
| \% | C | - Vector of const |  | vary |
| \% | L_initial | - Initial halfspan | ength | ft |
| \% | Lift_initial | - Initial lift |  | lb |
| \% | R_initial | - Initial location | f lift as fraction of halfspan | none |
| \% | tc_initial | - Initial thicknes | to chord ratio | none |
| \% | Twist_initial | - Initial wing twi |  | p.f. |
| \% | x_initial | - Initial wingbox | x-sectional thickness | p.f. |
| \% | X 1 (1) | - Wing taper ratio |  | none |
| \% | X1(2) | - Wingbox x -sec | onal area as poly. funct. | p.f. |
| \% | Y21 | - Lift |  | lb |
| \% | Y31 | - Engine weight |  | lb |
| \% | Z(1) | - Thickness/chor | ratio | none |
| \% | Z(2) | - Altitude |  |  |
| \% | Z(3) | - Mach number |  | none |
| \% | Z(4) | - Aspect ratio |  | none |
| \% | Z(5) | - Wing sweep |  | deg |
| \% | Z(6) | - Wing surface a |  | $\mathrm{ft}^{2}$ |
| \% ${ }^{\text {c }}$ |  |  |  |  |
| \% | Output Variables |  |  |  |
| \% | G1(1) | - Stress on wing |  | p.f. |
| \% | G1(2) | - Stress on wing |  | p.f. |
| \% | G1(3) | - Stress on wing |  | p.f. |

\%-----Preserve initial values for polynomial calculations-----\%
Linitial $=s q r t(10(8) * i 0(10) / 1+i$
R_initial $=(1+2 * i 0(1)) /(3 *(1+i 0(1)))$;
ESF_initial=Y32(1);
Cf_initial=i0(3);
Lift_initial=Y21(1);
M_initial=i0(7);
h_initial $^{2} \mathrm{i} 0(6)$;
T initial $=\mathrm{i} 0(4)$;

## \%-----Execute Gauss Seidel iteration on system to find Y variables-.---\%

$\mathrm{Lu}=\mathrm{Y} 21(1)+10 ;$

Y31(1) +10 ;
ESFu $=\mathrm{Y} 32(1)+10$;
while ((abs(Lu-Y21(1))>(Y21(1)*.001)) |(abs(Weu-Y31(1))>(Y31(1)*.001)) | (abs(ESFu-Y32(1))>(Y32(1)*.001)))
$\mathrm{Lu}=\mathrm{Y} 21(1) ;$
$\mathrm{Weu}=\mathrm{Y} 31(1) ;$
$\mathrm{ESFu}=\mathrm{Y} 32(1) ;$
ESFu=Y32(1);
\%-----Call Black Boxes-----\%
$[\mathrm{Y} 1, \mathrm{Y} 12, \mathrm{Y} 14, \mathrm{G} 1]=$
BB_weight(Z,Y21,Y31,X1,x_initial,L_initial,R_initial,Lift_initial,Twist_initial,tc_ini tial,C);
[Y2,Y21,Y23,Y24,G2]=BB_dragpolar(Z,Y12,Y32,X2,ESF_initial,Cf_initial,Twist_in itial,tc_initial,C);
[Y3,Y34,Y31,Y32,G3]=BB_power(Z,Y23,X3,M_initial,h_initial,T_initial,C);
$[\mathrm{Y} 4]=\mathrm{BB}$ _range(Z,Y14,Y24,Y34); [ Y 4 Y$]=\mathrm{BB}$ _range( $\mathrm{Z}, \mathrm{Y} 14, \mathrm{Y} 24, \mathrm{Y} 34$ );
\%-----Write post-iterative variable to output file-----\%



[^1]bound5 $=[.2,2,2,2, .2]$;
G1(3)=PolyApprox(S_initial5,S5,flag5,bound5); \%--wing stress
S_initial6=[tc_initial,Lift_initial,x_initial,L_initial,R_initial]; S6 $=[\mathrm{Z}(1), \mathrm{Y} 21(1), \mathrm{X} 1(2), \mathrm{L}, \mathrm{R}]$; flag $6=[4,1,4,1,1]$;
bound6 $=[.25, .25, .25, .25, .25]$;
Gl(4)=PolyApprox(S_initial6,S6,flag6,bound6); \%--wing stress
S_initial7=[tc_initial,Lift_initial,x_initial,L_initial,R_initial];
S7=[Z(1),Y21(1),X1(2),L,R];
flag $=[4,1,4,1,1]$;

$\mathrm{G} 1(6)=\mathrm{Y} 1(3) ; \%$--wing twist
\%---THIS SECTION COMPUTES CONSTRAINT POLYNOMIAL FUNCTIONS--\%
\% This subfunction calculates the drag and lift-to-drag ratio of the aircraft. It calls the subfunction POLYAPPROX to compute functions represented by polynomials.
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vary
p.f.
none
none
p.f.
p.f.
lb
p.f.


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- Vector of constants
- Initial coefficient of friction
- Initial engine scale factor
- Initial thickness to chord ratio

Initial wing twist

- Coefficient of friction
- Total aircraft weight
- Wing twist

$\mathrm{V}=\mathrm{Z}(3) * 968.1$; rho $=(2.377 \mathrm{e}-03)^{*}(.2971) * \exp (-(\mathrm{Z}(2)-36089) / 20806.7)$;
$\mathrm{CL}=\mathrm{Y} 12(1) /\left(.5 *\right.$ rho $\left.{ }^{*}\left(\mathrm{~V}^{\wedge} 2\right) * \mathrm{Z}(6)\right) ; \%$-Lift Coefficient
\%-----Polynomial function modifying CDmin for ESF and friction coefficient-----\%
S_initial1=[ESF_initial,Cf_initial];
S1=[Y32(1),X2(1)];
flag $1=[1,1] ;$ 25];
bound $1=[.25, .25]$,
Fol $=$ PolyApprox $($ S_initial 1, S1,flag 1,bound 1$)$; .
if $Z(3)>=1$
$\mathrm{k}=\mathrm{Z}(4)^{*}\left(\mathrm{Z}(3)^{\wedge} 2-1\right)^{*} \cos \left(\mathrm{Z}(5)^{*} \mathrm{p} / 180\right) /\left(4 * \mathrm{Z}(4)^{*} \operatorname{sqrt}\left(\mathrm{Z}(5)^{\wedge} 2-1\right)-2\right) ;$
$\mathrm{k}=1 /\left(\mathrm{pi}{ }^{*} 0.8^{*} \mathrm{Z}(4)\right)$;

S_initial2=[Twist_initial];
S2 $=[\mathrm{Y} 12(2)]$;
flag2=[5];
bound2 $=[.25]$;
Fo2=PolyApprox(S_initial2,S2,flag2,bound2);
$\mathrm{CD}=\mathrm{Fo} 2^{*}\left(\mathrm{CDmin}+\mathrm{k}^{*}\left(\mathrm{CL}^{\wedge} 2\right)\right.$;




| \% | G3(3) | - Throttle setting constraint |
| :---: | :---: | :---: |
| \% | Y3(1) | - Specific fuel consumption |
| \% | Y3(2) | - Engine weight |
| \% | Y3(3) | - Engine scale factor |
| \% | Y31 | - Engine weight |
| \% | Y32 | - Engine scale factor |
| \% | Y34 | - Specific fuel consumption |
| \% |  |  |
| \% | Local Variables | : |
| \% | Dim_Throttle | - Non-dimensional throttle setting |
| \% | p | - Vector of constant coefficients for upper |
| \% |  | limit on throttle setting surface fit |
| \% | s | - Vector of constant coefficients for SFC |
| \% |  | surface fit |
| \% | Thrust | - Thrust required |
| \% | Throttle_uA | - Upper limit on throttle setting |
| \% |  |  |
| \% | Subfunctions | : |
|  | PolyApprox | -Forms polynomial functions for |
|  | \% |  |
|  |  |  |




| \% | Y24 | - Lift-to-drag ratio | none |
| :---: | :---: | :---: | :---: |
| \% | Y34 | - Specific fuel consumption | 1/hr |
| \% | Z(1) | - Thickness/chord ratio | none |
| \% | Z(2) | - Altitude | ft |
| \% | Z(3) | - Mach number | none |
| \% | Z(4) | - Aspect ratio | none |
| \% | Z(5) | - Wing sweep | deg |
| \% | Z(6) | - Wing surface area | $\mathrm{ft}^{2}$ |
| \% |  |  |  |
| \% Output Variables |  | : | NM |
| \% | Y4 | - Range |  |
| \% |  |  |  |
| \% Local Variables |  | : |  |
| \% Theta |  | - Temperature ratio | none |
|  |  |  |  |
| $\%-$ |  |  |  |
| $\text { function }[\mathrm{Y} 4]=\mathrm{BB} \text { _range }(\mathrm{Z}, \mathrm{Y} 14, \mathrm{Y} 24, \mathrm{Y} 34)$ |  |  |  |
| \% -----THIS SECTION COMPUTES THE A/C RANGE (Breguet)-----\% |  |  |  |
| $\begin{aligned} & \text { if } Z(2)<36089 \\ & \text { theta }=1-0.000006875 * Z(2) \end{aligned}$ |  |  |  |
| el | theta=.7519; |  |  |
| $\mathrm{Y} 4(1)=\left((\mathrm{Z}(3) * \mathrm{Y} 24(1))^{*} 661 *\right.$ sqrt(theta)/Y34(1))* $\log (\mathrm{Y} 14(1) /(\mathrm{Y} 14(1)-\mathrm{Y} 14(2))$ ); |  |  |  |
| \%-----THIS SECTION COMPUTES THE A/C RANGE-----\% |  |  |  |
| \% |  |  |  |
|  |  |  |  |  |  |
| \% |  | Subfunction POLYAPPROX |  |
| \% |  |  |  |
| \% | This subfunction | culates polynomial coefficient | beha |
|  | of certain synthe | variables and function modifie | each |
|  | polynomial are ser | ted based on knowledge of ea | fier's |

$\quad \mathrm{b}=2 * \mathrm{a}$;
end
\%DETERMINE BOUNDS ON FF DEPENDING ON SLOPE-SHAPE
\%CALCULATE POLYNOMIAL COEFFICIENTS (S-ABOUT ORIGIN)



end
end
\%FILL UP OFF DIAGONALS OF Aij
$\begin{array}{ccccc}\mathrm{R}=[\ldots & & & & \\ 0.2736 & 0.3970 & 0.8152 & 0.9230 & 0.1108 \\ 0.4252 & 0.4415 & 0.6357 & 0.7435 & 0.1138 \\ 0.0329 & 0.8856 & 0.8390 & 0.3657 & 0.0019 \\ 0.0878 & 0.7248 & 0.1978 & 0.0200 & 0.0169 \\ 0.8955 & 0.4568 & 0.8075 & 0.9239 & 0.2525] ;\end{array}$
for $\mathrm{i}=1: \operatorname{size}(\mathrm{S}, 2)$
for $\mathrm{j}=(\mathrm{i}+1): \operatorname{size}(\mathbf{S}, 2)$
$\quad \operatorname{Aij(i,j)}=\operatorname{Aij(i,i)} \boldsymbol{*} \times(\mathrm{i}, \mathrm{j}) ;$
$\quad \operatorname{Aij}(\mathrm{j}, \mathrm{i})=\operatorname{Aij}(\mathrm{i}, \mathrm{j}) ;$
end
end
\%CALCULATE POLYNOMIAL
$\mathrm{FF}=\mathrm{Ao}+\mathrm{Ai}^{*}\left(\mathrm{~S}_{\mathrm{s}}\right.$ shifted $)+(1 / 2)^{*}$
$\mathrm{FF}=\mathrm{Ao}+\mathrm{Ai}^{*}\left(\mathrm{~S}_{-}\right.$shifted $)+(1 / 2)^{*}\left(\mathrm{~S}_{\_} \text {shifted }\right)^{*}\left(\mathrm{Aij}^{\prime}\right)^{*}\left(\mathrm{~S}_{\text {_shifted }}{ }^{\prime}\right) ;$
function [FF, Ao, Ai, Aij] = PolyApprox(S, S_new, flag, S_bound)
for $\mathrm{i}=1$ :size( $\mathrm{S}, 2$ )
S_norm(i) = S_new(i)/S(i); \%normalize new $S$ with initial $S$ if S_norm(i)>1.25

S_norm(i) $=1.25$;
elseif $S$ _norm(i) $<0.75$
S_norm $(\mathrm{i})=0.75$;
S_shifted(i) = S_norm(i) - 1; \%shift S vector near origin

if flag $(i)==5$
\%CALCULATE POLYNOMIAL COEFFICIENTS (S-ABOUT ORIGIN)
So=0;
$\mathrm{Sl}=\mathrm{So}-\mathrm{S}$ _bound $(\mathrm{i}) ;$
$\mathrm{Su}=\mathrm{So}+\mathrm{S}$ bound $(\mathrm{i}) ;$
Mtx_shifted $=\left[1 \mathrm{SI}^{\mathrm{S}}{ }^{\wedge} 2 ; 1\right.$ So $\left.\mathrm{So}^{\wedge} 2 ; 1 \mathrm{Su} \mathrm{Su}{ }^{\wedge} 2\right]$; $F_{\text {_bound }}=\left[1+\left(.5^{*} \mathrm{a}\right)^{\wedge} 2 ; 1 ; 1+\left(.5^{*} \mathrm{~b}\right)^{\wedge} 2\right]$;
A = Mtx_shiftedIF_bound;
A $=$ Mtx_shiftedF_bound;
Ao $=\mathbf{A}(1) ;$
$\mathrm{Ao}=\mathrm{A}(1) ;$
$\mathrm{Ai}(\mathrm{i})=\mathrm{A}(2) ;$

\%CALCULATE POLYNOMIAL COEFFICIENTS
else





fun-
tion[dX3,Lagrange3,fstore,gstore0]=BB3OPT(vlb_nd,vub_nd,i0, P_var,Z,Y23,X3,M_i nitial,h_initial,T_initial,C,DY_T)
options(1)=1;
options(2)=.0001;
options(3)=.0001;
options(4)=.001;
options(14)=1000;
[fstore0,gstore0,dX30]=BB3WRAPPER(x0,i0,P_var,Z,Y23,M_initial,h_initial,T_initi al,C,DY_T);
[x,options,Lagrange3]=constr('BB3WRAPPER',x0,options,vlb,vub,[],i0,P_var,Z,Y23, M_initial,h_initial,T_initial,C,DY_T);
[fstore,gstore,dX3]=BB3WRAPPER(x,i0,P_var,Z,Y23,M_initial,h_initial,T_initial,C, DY_T);

| \% |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |
| \% | Subfunction BB3WRAPPER |  |  |  |
| \% |  |  |  |  |
| \% | This subfunction computes the objective function and the constraints for the |  |  |  |
| \% | local optimization on the POWER module. |  |  |  |
| \% |  |  |  |  |
| \% | Author | : Jeremy S. Agte | NASA Langley/GWU | Spring '98 |
| \% |  |  |  |  |
| \% | Input Variables | : |  |  |
| \% | C | - Vector of cons |  | vary |
| \% | DY_T | - Vector of total | erivatives, behavior |  |
| \% |  | variables w.r.t | rottle setting | vary |
| \% | h_initial | - Initial altitude |  | ft |

vary
none
 - see Matlab 'constr' function


$$
\begin{array}{lc}
\text { - Vector of optimal changes in X3 variables } & \text { vary } \\
\text { - Value of objective function for BB3 optim. } & \text { NM } \\
\text { - Vector of local constraint values at beginning } & \\
\begin{array}{ll}
\text { of BB3 optimization }
\end{array} & \text { p.f. } \\
\begin{array}{l}
\text { Vector of Lagrange multipliers from BB3 at }
\end{array} & \\
\quad \text { optimum }
\end{array}
$$ Non-dimensional lower bounds on BB3 design none


을 none Non-dimensional upper bounds on BB3 design variables Vector of non-dimensional starting points for BB3 optimization
 for BB3 optimization
Matlab optimization routine
P_var
T_initial
vlb_nd
vub_nd

X3
$Y 23$
$Z(1)$
$Z(2)$
$Z(3)$
$Z(4)$
$Z(5)$
$Z(6)$
Output Variables
dX3
fstore
gstore 0 Local Variables options
vlb
Lagrange3

vub
x 0

| T_initial <br> vlb_nd |
| :---: |
| vub_nd |
| X3 |
| Y23 |
| Z(1) |
| Z(2) |
| Z(3) |
| Z(4) |
| Z(5) |
| Z(6) |
| Output Variables <br> dX3 <br> fstore <br> gstore0 |
| Lagrange3 |
| Local Variables options vlb |
| vub |
| x 0 |
| Subfunctions |
| BB3WRA |
| constr |

$$
\begin{aligned}
& \mathrm{vlb}=\left[\mathrm{vlb} \_ \text {nd(4) }\right] ; \\
& \mathrm{vub}=\left[\mathrm{vub} \_ \text {nd(4) }\right]
\end{aligned}
$$

$$
\mathrm{x} 0=[\mathrm{X} 3(1) / \mathrm{i} 0(4)-1]
$$



| \% | dZ - V | - Vector of optimal changes in Z variables | vary |
| :---: | :---: | :---: | :---: |
| \% | fstore - V | - Value of objective function for system optim. | NM |
| \% | gstore0 - | - Vector of constraint values at beginning of system optimization | vary |
| \% |  |  |  |
| \% | Local Variables | : |  |
| \% | options - s | - see Matlab 'constr' function |  |
| \% | vlb - | - Non-dimensional lower bounds on Z |  |
| \% |  | variables | none |
| \% | vub - | - Non-dimensional upper bounds on Z |  |
| \% |  | variables | none |
| \% | x 0 - | - Vector of non-dimensional starting points |  |
| \% |  | for system optimization | none |
| \% |  |  |  |
| \% | Subfunctions | : |  |
| \% | SysWRAPPER | R - Contains objective function and constraints |  |
| \% |  | for system optimization |  |
| \% |  | - Matlab optimization routine |  |
| \% | constr |  |  |
| \% |  |  |  |


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 $d Y_{-}$Sref
$d Y_{-} T$
$d Y_{-} t c$ dY_x - Coefficient matrix in GSE

- Vector of derivatives, BB1 - Vector of derivatives, BB1 constaints
w.r.t Z variables - Vector of derivatives, BB2 constaints w.r.t $Z$ variables - Vector of derivatives, BB3 constaints w.r.t Z variables - Vector of derivatives, BB1 constaints - Vector of derivatives, BB2 constaints - Vector of derivatives, BB3 constaints w.r.t Y variables entering BB3 Vector of partial derivatives, behavior variables w.r.t aspect ratio
- Vector of partial derivatives, behavior variables w.r.t skin friction coefficient
 variables w.r.t altitude
- Vector of partial derivatives, behavior variables w.r.t wing sweep
- Vector of partial derivatives, behavior variables w.r.t taper ratio
- Vector of partial derivatives, behavior
variables w.r.t Mach number
- Vector of partial derivatives, behavior
variables w.r.t wing surface area
- Vector of partial derivatives, behavior
variables w.r.t throttle setting
- Vector of partial derivatives, behavior
variables w.r.t thickness/chord ratio
- Vector of partial derivatives, behavior

Altitude



 $\begin{array}{ll}\text { G1(6) } & \text { - Wing twist as constraint } \\ \text { G2 } & \text { - Pressure gradient } \\ \text { G3(1) } & \text { - Engine scale factor constraint } \\ \text { G3(2) } & \text { - Engine temperature } \\ \text { G3(3) } & \text { - Throttle setting constraint } \\ \text { h_initial } & \text { - Initial altitude } \\ \text { L_initial } & \text { - Initial halfspan length } \\ \text { Lift_initial } & \text { - Initial lift } \\ \text { M_initial } & \text { - Initial Mach number } \\ \text { R_initial } & \text { - Initial location of lift as fraction of halfspan } \\ \text { tc_initial } & \text { - Initial thickness to chord ratio } \\ \text { T_initial } & \text { - Initial throttle setting } \\ \text { Twist_initial } & \text { - Initial wing twist } \\ \text { X_initial } & \text { - Initial wingbox x-sectional thickness } \\ \text { Y1(1) } & \text { - Total aircraft weight } \\ \text { Y1(2) } & \text { - Fuel weight } \\ \text { Y1(3) } & \text { - Wing twist } \\ \text { Y12(1) } & \text { - Total aircraft weight } \\ \text { Y12(2) } & \text { - Wing twist } \\ \text { Y14(1) } & \text { - Total aircraft weight } \\ \text { Y14(2) } & \text { - Fuel weight } \\ \text { Y2(1) } & \text { - Lift } \\ \text { Y2(2) } & \text { - Drag } \\ \text { Y2(3) } & \text { - Lift-to-drag ratio } \\ \text { Y21 } & \text { - Lift } \\ \text { Y23 } & \text { - Drag } \\ \text { Y24 } & \text { - Lift-to-drag ratio } \\ \text { Y3(1) } & \text { - Specific fuel consumption } \\ \text { Y3(2) } & \text { - Engine weight } \\ \text { Y3(3) } & \text { - Engine scale factor } \\ \text { Y31 } & \text { - Engine weight } \\ \text { Y32 } & \text { - Engine scale factor } \\ \text { Y34 } & \text { - Specific fuel consumption } \\ \text { Y4 } & \text { - Objective function output from BB4 } \\ \text { X1(1) } & \text { - Wing taper ratio } \\ \text { X1(2) } & \text { - Wingbox x-sectional area as poly. funct. } \\ \text { X2 } & \text { - Skin friction coefficient as poly. funct. } \\ \text { X3 } & \text { - Throttle setting } \\ \text { Z(1) } & \text { - Thickness/chord ratio } \\ & \end{array}$


func-
tion[A,dY_lambda,dY_x,dY_Cf,dY_T,dY_tc,dY_h,dY_M,dY_AR,dY_Lambda,dY_S
ref,dg1_Z,dg2_Z,dg3_Z,dg1_YE1,dg2_YE2,dg3_YE3]=FIN_DIFF(Z,Y1,Y2,Y3,Y4,
Y12,Y14,Y21,Y23,Y24,Y31,Y32,Y34,X1,X2,X3,G1,G2,G3,C,Twist_initial,x_initial,
L_initial,R_initial,ESF_initial,Cf_initial,Lift_initial,tc_initial,M_initial,h_initial,T_in
itial)
\%-----calculate Y partials-...-\%
[A12]=fin_diff_A12(Z,Y1,Y21,Y31,X1,C,x_initial,L_initial,R_initial,Lift_initial,Twi
st_initial,tc_initial);
[A13]=fin_diff_A13(Z,Y1,Y21,Y31,X1,C,x_initial,L_initial,R_initial,Lift_initial,Twi
st_initial,tc_initial);
[A21]=fin_diff_A21(Z,Y2,Y12,Y32,X2,C,Twist_initial,ESF_initial,Cf_initial,tc_initi
al);
[A23]=fin_diff_A23(Z,Y2,Y12,Y32,X2,C,Twist_initial,ESF_initial,Cf_initial,tc_initi
al);
(A32]=fin_diff_A32(Z,Y3,Y23,X3,C,M_initial,h_initial,T_initial);
[A41]=fin_diff_A41(Z,Y4,Y14,Y24,Y34);
[A42]=fin_diff_A42(Z,Y4,Y14,Y24,Y34);
[A43]=fin_diff_A43(Z,Y4,Y14,Y24,Y34);
[dg1_Y21]=fdG1_Y21(Z,Y1,Y21,Y31,X1,G1,C,x_initial,L_initial,R_initial,Lift_initi al,Twist_initial,tc_initial);
[dg1_Y31]=fdG1_Y31(Z,Y1,Y21,Y31,X1,G1,C,x_initial,L_initial,R_initial,Lift_initi
al,Twist_initial,tc_initial);
$\mathrm{dg} 1 \_\mathrm{YE} 1=\left[\mathrm{dg} 1 \_\mathrm{Y} 21 \mathrm{dg} 1 \_\mathrm{Y} 31\right] ;$
[dg2_Y12]=fdG2_Y12(Z,Y2,Y12,Y32,X2,G2,C,Twist_initial,ESF_initial,Cf_initial,t c initial);
[dg2_Y32]=fdG2_Y32(Z,Y2,Y12,Y32,X2,G2,C,Twist_initial,ESF_initial,Cf_initial,t c_initial);
dg2_YE2 = [dg2_Y12 dg2_Y32];
$\left[d g 3 \_Y 23\right]=f d G 3 \_Y 23(Z, Y 3, Y 23$, dg3_YE3 = [dg3_Y23];
㖘



## \%-----calculate Z partials-----\%

[dY1_Z1,dY1_Z4,dY1_Z5,dY1_Z6]=fdY1_Z(Z,Y1,Y21,Y31,X1,C,x_initial,L_initial,

$$
\begin{aligned}
& \text { [dY1_X1_1,dY1_X1_2]=fdY1_X1(Z,Y1,Y21,Y31,X1,C,x_initial,L_initial,R_initial, } \\
& \text { Lift_initial,Twist_initial,tc_initial); } \\
& \text { [dY2_X2]=fdY2_X2(Z,Y2,Y12,Y32,X2,C,Twist_initial,ESF_initial,Cf_initial,tc_initi } \\
& \text { al); } \\
& \text { [dY3_X3]=fdY3_X3(Z,Y3,Y23,X3,C,M_initial,h_initial,T_initial); }
\end{aligned}
$$

## -A41(1,1)-A41(1,2) 000 -A42(1,1)-A43(1,1) 00 1];

## \%-----calculate X partials-----\%

[dg1_Z]=fdG1_Z(Z,Y1,Y21,Y31,X1,G1,C,x_initial,L_initial,R_initial,Lift_initial,Tw ist_initial.tc_initial);
[dg2_Z]=fdG2_Z(Z,Y2,Y12,Y32,X2,G2,C,Twist_initial,ESF_initial,Cf_initial,tc_initi
[dg3_Z] $=\mathrm{fdG} 3 \_Z\left(Z, Y 3, Y 23, X 3, G 3, C, M \_i n i t i a l, h \_i n i t i a l, T \_i n i t i a l\right) ; ~$

[^2]


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    ${ }^{\ddagger}$ Professor, George Washington University, Joint Institute for Advancement of Flight Sciences, AIAA Fellow

[^1]:    func-
    tion[Y2,Y21,Y23,Y24,G2]=BB_dragpolar(Z,Y12,Y32,X2,ESF_initial,Cf_initial,Twis
    t_initial,tc_initial,C)
    \%----つ/V 3HL HO DV\&G TVLOL THL SHLRdWOP NOILכヨS SIHL----\%
    $\mathrm{V}=\mathrm{Z}(3)^{*}(1116.39 * \operatorname{sqrt}(1-(6.875 \mathrm{e}-06 * \mathrm{Z}(2)))) ;$
    rho $=(2.377 \mathrm{e}-03)^{*}(1-(6.875 \mathrm{e}-06 * \mathrm{Z}(2)))^{\wedge} 4.2561 ;$

[^2]:    dY_lambda $=\left[d Y 1 \_X 1 \_1 ; 0 ; 0 ; 0 ; 0 ; 0 ; 0 ; 0\right] ;$ dY_x = [dY1_X1_2; 0; 0; 0; 0;0;0;0];

