NASA/TM-1998-208715



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August 1998

Acknowledgments

The authors gratefully acknowledge the contributions of Dr. Srinivas Kodiyalam of the Engineous Co. for providing test results for the Aircraft Optimization and the Electronic Package Optimization using a software package called iSIGHT.

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BI-LEVEL INTEGRATED SYSTEM SYNTHESIS (BLISS)

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<u>Abstract</u>

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.

<u>0. Introduction</u>

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

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 Sobieszczanski-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 (Sobieszczanski-Sobieski, 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

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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. Because 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.
- $BBA(Y_r,(Z,X_r))$ analysis of BB_r to compute Y_r for given Z and X_r
- $BBOF_r$ BB Objective Function computed in BB_r BBOPT_r(X_r, ϕ_r ,G_r) - optimization in BB_r defined by
- eq.(2.1/9) BBOSA_r($X_{r,opt}, Z, Y_{r,s}$) - analysis of BB optimum for

sensitivity to parameters

- BBSA(D(Y_r ,(Z,X_r, $Y_{r,s}$)) sensitivity analysis of BB_r to compute its output derivatives w.r.t. Z, X_r, and $Y_{r,s}$
- D(V1,V2) total derivative dV1/dV2
- d(V1,V2) partial derivative $\partial V1/\partial V2$; 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 NG_o
- G_r vector of the constraint functions, $g_{r,t}$ local to BB_r , $g_{r,t} \leq 0$ is a satisfied constraint
- G_{yz} constraints in a BB that have a stronger depend ence on Y and Z, than on X
- GSE Global Sensitivity Equations (Sobieszczanski-Sobieski, 1990); GSE/OS - GSE/Optimized Subsystems.
- I identity matrix.
- L vector of the Lagrange multipliers corresponding to G_o , length NG_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, Φ) system objective optimization defined by eq. (2.2.3/1)
- SSA(D(Y,(Z,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 NX_r, these variables are local to BB_r; X without subscript - a vector of all concatenated X_r, length NX
- XL, XU lower and upper bounds on X, sideconstraints.
- Y_r vector of behavior (state) variables output from BB_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 Y_r length is NY_r; Y without subscripts - a vector of all concatenated Y_r , length NY

- $Y_{r,s}$ vector of variables input to BB_r from BB_s, these are the coupling variables; an element of $Y_{r,s}$ is denoted $y_{r,s,i}$; note that by this definition $Y_{r,s}$ is a subset of $Y_{r,s}$ vector length $NY_{r,s}$
- Z vector of the design variables z_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.
- ZL, ZU lower and upper bounds on Z, sideconstraints

 Δ - increment

 ΔZL , ΔZU - move limits

- ϕ_r the local objective function in BB_r
- Φ 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.

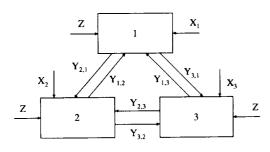


Figure 1: System of Coupled BBs

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
- Φ maximum range for given mission characteristics
- $\begin{array}{l} Y_{1,2} \mbox{- includes the aerodynamic drag; } Y_{1,3} \mbox{- includes the structural weight; } Y_{2,1} \mbox{- includes Mach number; } Y_{3,1} \mbox{- includes TOGW; } Y_{2,3} \mbox{- includes the structural deformations that alter the aerodynamic shape; } Y_{3,2} \mbox{- includes the aerodynamic loads } \end{array}$
- g1,t a noise abatement constraint on the mission pro-

file; $g_{2,t}$ - limit of the chordwise pressure gradient; $g_{3,t}$ - allowable stress

- x_{1,j} cruise altitude; x_{2,j} leading edge radius; x_{3,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:	X and Z	(1)
Find:	ΔX and ΔZ	
Minimize:	$\Phi(X,\!Z,\!Y(X,\!Z))$	
Satisfy:	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 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 BB_1 so that

$$\mathbf{\Phi} = \mathbf{y}_{1,i} \tag{1}$$

is one of the elements of Y_1 .

Total derivatives of Y w.r.t. $x_{r,j}$, D(Y, $x_{r,j}$), 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 $x_{r,i}$

$$[A] \{ D(Y, x_{r,j}) \} = \{ d(Y, x_{r,j}) \}$$
(2)

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

$$\begin{bmatrix} I & A_{1,2} & A_{1,3} \\ \hline A_{2,1} & I & A_{2,3} \\ \hline A_{3,1} & A_{3,2} & I \end{bmatrix}$$
(3)

where I stands for identity matrix, $NY_r xNY_r$, and $A_{r,s}$ are matrices of the derivatives that capture sensitivity of the BB_r output to input. For example

$$A_{2,3} = -[d(Y_2, Y_3)], NY_2 x NY_3$$

$$A_{3,2} = -[d(Y_3, Y_2)], NY_3 x NY_2$$
(4)

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

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

$$\Phi = y_{1,i} = (y_{1,i})_0 + D(y_{1,i}, X_1)^T \Delta X_1 + D(y_{1,i}, X_2)^T \Delta X_2 + D(y_{1,i}, X_3)^T \Delta X_3$$
(5)

where D-terms are vectors of length NX_r.

We see from eq. 5 that

$$\Delta \Phi = D(y_{1,i}, X_1)^T \Delta X_1 + D(y_{1,i}, X_2)^T \Delta X_2 + D(y_{1,i}, X_3)^T \Delta X_3$$
(6)

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 BB₂ as an example, objective ϕ_2 is

$$\phi_2 = D(y_{1,i}, X_2)^T \Delta X_{2,j}, j = 1 - -> N X_2$$
 (7)

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 BB_r design variables X_r on the entire system objective function.

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

$$\phi_2 = D(y_{1,i}, x_{2,1})^T \Delta X_{2,1} + D(y_{1,i}, x_{2,2})^T \Delta X_{2,2} + \dots + D(y_{1,i}, x_{2,j})^T \Delta X_{2,j} + \dots, j = 1 - -> NX_2$$
(8)

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 BB_r where Φ is being computed. In the aircraft example it is $\Phi = y_{1,i}$ in BB₁ according to eq. 1. However, the BB₁ optimization objective is not $\phi_1 = y_{1,i}$. Instead, it is ϕ_1 from an equation analogous to eq.8.

The local optimization problem may be stated formally for BB₂

Given: X_2 , Z, and $Y_{2,1}$, $Y_{2,3}$ (9)Find: ΔX_2 ; length NX_2 Minimize: $\phi_2 = D(y_{1,i}, X_2)^T \Delta X_2$ Satisfy: $G_2 \le 0$, 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 BB_1 , and 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 Φ 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(y_{1,i},Z).

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 BB as parameters. The modification leads to a new generalization of GSE that takes the following form

$$\begin{bmatrix} \mathbf{M} \end{bmatrix} \begin{cases} \mathbf{D}(\mathbf{Y}, \mathbf{z}_k) \\ \mathbf{D}(\mathbf{X}, \mathbf{z}_k) \end{cases} = \begin{cases} \mathbf{d}(\mathbf{Y}, \mathbf{z}_k) \\ \mathbf{d}(\mathbf{X}, \mathbf{z}_k) \end{cases}$$
(1)

termed GSE/OS for GSE/Optimized Subsystems. The GSE/OS yields a vector $D(Y,z_k)$ and $D(X,z_k)$, and because Φ is one of the elements of Y, $\Phi = y_{1,i}$, we get the desired derivative $D(\Phi,z_k)$. 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 $d(Y_r, Y_s)$, $d(Y_r, X_r)$, and $d(X_r, Y_s)$. These terms and the RHS terms of $d(Y, z_k)$ and $d(X, z_k)$ are obtained from the following sources

• d(Y_r,Y_s), d(Y_r,X_r), and d(Y,z_k) ----- from BBSA • d(X_r,z_k), d(X_r,Y_s) ----- from BBOSA The terms $d(X,z_k)$ and $d(X_r,Y_s)$ 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 Sobieszczanski-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 $d(X_r, z_k)$ and $d(X_r, Y_s)$ 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 Φ , for the constraint changes caused by incrementing p_i . For a general case of the objective F=F(P) and G₀=G₀(P), the algorithm gives the following formula for D(F,P)₀

$$D(F,P)_o = d(F,P) + L^T d(G_o,P)$$

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

$$D(y_{1,i},Z)_{o}^{T} = (L^{T}d(G_{o},Z))_{1} + (L^{T}d(G_{o},Z))_{2} + (L^{T}d(G_{o},Z))_{3} + [(L^{T}d(G_{o},Y))_{1} + (L^{T}d(G_{o},Y))_{2} + (L^{T}d(G_{o},Y))_{3}](D(Y,Z)) + D(y_{1,i},Z)^{T}$$
(1)

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:

- $d(G_o,Z)$ and $d(G_o,Y)$ BBSA performed on isolated BB_r
- L obtained for BB_r at the end of BBOPT
- D(Y,Z) from GSE in SSA
- $D(y_{1,i},Z)$ the column corresponding to $y_{1,i}$ in the above matrix D(Y,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 $D(y_{1,i},Z)$ have been computed from either eq.(2.2.1/1) or as $D(y_{1,i},Z)_o$ from eq. (2.2.2/1), we can further improve the system objective by executing the following optimization, using any suitable optimizer

Given:Z and Φ_0 (1)Find: ΔZ Minimize: $\Phi = \Phi_0 + D(y_{1,i},Z)^T \Delta Z$ Satisfy: $ZL \leq Z + \Delta Z \leq ZU; \Delta ZL \leq \Delta Z \leq \Delta ZU$

Where Φ_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 D(y_{1,i}, Z) is a constrained derivative that protects $G_0 = 0$ 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_{yz} , 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

$$G_{yz}^{T} = G_{yz,0}^{T} + (d(G_{yz},Z) + d(G_{yz},Y)D(Y,Z))^{T}\Delta Z \le 0$$
(2)

where $d(G_{yz},Z)$, and $d(G_{yz},Y)$ 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.

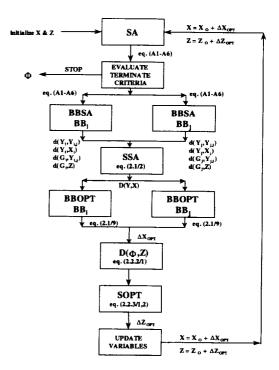


Figure 2: BLISS/B Flowchart

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 X & 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), $d(Y_{r,s},Y_s)$, d(G,Z), and d(G,Y), and SSA, eq. (2.1/2), to get D(Y,X) [and

D(Y,Z)]; Here is an opportunity for concurrent processing.

4. BBOPT for all BBs, eq. (2.1/9) using ϕ formulated individually for each BB (eq. (2.1/6, 7)), get ϕ_{opt} and ΔX_{opt} ; obtain L for G_o [skip L]. <u>Here is an opportunity for concurrent processing</u>.

5. Obtain $D(\Phi,Z)$ 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 ΔZ_{opt} by eq. (2.2.3/1 and 2) herein.

7. Update all quantities, and repeat from 1.

$$X = X_0 + \Delta X_{opt}; Z = Z + \Delta Z_{opt}$$

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.

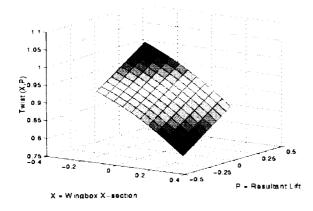


Figure 3: Polynomial Representation of Wing Twist

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 domain of NB X-spaces (Step #4) and the single Z-space (Step #6).

Caveat: because in BLISS/B the extrapolation of Φ in eq. (2.2.3/1) is based on the Lagrange multipliers in eq. (2.2.2/1), its accuracy depends on the BBOPT vielding a feasible solution, and on the active constraints Go remaining active for updated Z. If some constraints leave the active set G_o, 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 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 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.

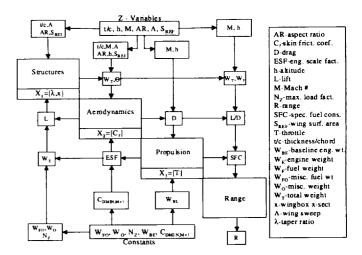


Figure 4: Data Dependencies for Range Optimization

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 (BB₁), aerodynamics (BB₂), propulsion (BB₃), and aircraft range (BB₄). All the disciplines were represented by modules comprising an analysis level typical for an early conceptual design stage.

var \ cycle*	1	2	3	4	5
Range (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
BB2 Extpl.	16.85	0.00	0.00	0.00	0.00
BB3 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
λ	0.25	0.14951	0.17476	0.25775	0.38757
X	1	0.75	0.75	0.75	0.75
С,	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
^(°)	55	66	70	70	70
S _{ref} (ft ²)	1000	1200	1400	1500	1500

*One cycle is one pass through the BLISS procedure

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 BB₂ 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	λ	x	С,	т	Vc
W _T	0.01146	1.71536	0.01981	-0.15744	0.12714
₩ŗ	0	0	0	0	0.72626
θ	-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
L/D	0.0115	1.7095	-0.1046	-0.15694	-0.54935
SFC	1.98E-20	-5.07E-18	-2.70E-17	0.08544	0
Wr	-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	Sref
num \ den Wτ	h -0.33931	M 0.31958	AR 0.08208	Sweep 0.2537	S _{ref} 0.55182
Wτ	-0.33931	0.31958	0.08208	0.2537	0.55182
W _T W _F	-0.33931 0	0.31958 0	0.08208	0.2537 0	0.55182
W _T W _F Ø	-0.33931 0 -1.93E-13	0.31958 0 -6.15E-14	0.08208 -0.36043 -0.10766	0.2537 0 3.77E-14	0.55182 1.09211 -0.10766
Wr Wr e L	-0.33931 0 -1.93E-13 -0.33931	0.31958 0 -6.15E-14 0.31958	0.08208 -0.36043 -0.10766 0.08208	0.2537 0 3.77E-14 0.2537	0.55182 1.09211 -0.10766 0.55182
₩ ₇ ₩₅ Θ L D	-0.33931 0 -1.93E-13 -0.33931 -2.1339	0.31958 0 -6.15E-14 0.31958 2.00984	0.08208 -0.36043 -0.10766 0.08208 3.37E-06	0.2537 0 3.77E-14 0.2537 -0.83983	0.55182 1.09211 -0.10766 0.55182 0.99638
₩ ₇ ₩₅ € L D L/D	-0.33931 0 -1.93E-13 -0.33931 -2.1339 1.84108	0.31958 0 -6.15E-14 0.31958 2.00984 -1.6507	0.08208 -0.36043 -0.10766 0.08208 3.37E-06 0.08207	0.2537 0 3.77E-14 0.2537 -0.83983 1.10064	0.55182 1.09211 -0.10766 0.55182 0.99638
W _T W _F O L D L/D SFC	-0.33931 0 -1.93E-13 -0.33931 -2.1339 1.84108 0.12946	0.31958 0 -6.15E-14 0.31958 2.00984 -1.6507 0.05555	0.08208 -0.36043 -0.10766 0.08208 3.37E-06 0.08207 2.31E-17	0.2537 0 3.77E-14 0.2537 -0.83983 1.10064 -1.86E-16	0.55182 1.09211 -0.10766 0.55182 0.99838 -0.43675 0

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 BB_4 is an analysis-only module and does not feedback any data to other BBs.

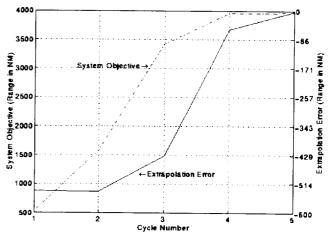


Figure 5: Range and Extrapolation Error Histogram

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.

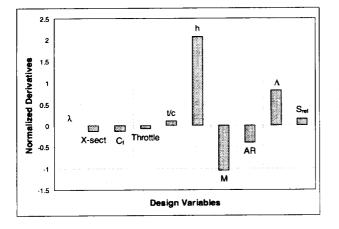


Figure 6: Range Sensitivities (1st cycle)

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, optimal t/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.

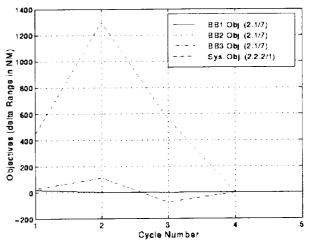


Figure 7: BB and System Contributions to Range

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.

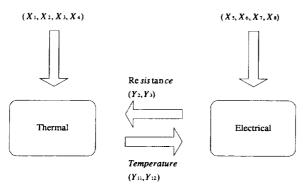


Figure 8: Electronic Packaging Data Dependencies

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.

		Initial Design	Init. Des. Max	Final Design	Fin. Des. Max	
Method	Case	Objective	Constr. Viol.	Objective	Constr. Viol.	Work
A-in-O	1	7.79440E+01	2.16630E-08	6.39720E+05	1.22E-03	498
	2	6.83630E+03	-2.89560E-01	6.39720E+05	1.22E-03	264
	3	1.51110E+03	-4.29240E-02	6.36540E+05	1.45E-03	264
	4	1.46070E+01	-1.02490E-03	6.36940E+05	1.42E-03	175
BLISS/A	1	7.79440E+01	2.16630E-08	6.39700E+05	1.20E-03	436
	2	6.83630E+03	-2.89560E-01	6.39050E+05	1.18E-03	508
	3	1.51110E+03	-4.29240E-02	6.39050E+05	-4.89E-04	174
	4	1.46070E+01	-1.02490E-03	6.39290E+05	3.70E-04	313
BLISS/B	1	7.79440E+01	2.16630E-08	6.39720E+05	1.22E-03	365
	2	6.83630E+03	-2.89560E-01	6.39720E+05	1.22E-03	207
	3	1.51110E+03	-4.29240E-02	6.39720E+05	1.22E-03	114
	4	1.46070E+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 gradient-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 $y_{r,s,i}$ - 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 Sobieszczanski-Sobieski, 1990. The condensed derivation is provided below. It begins by recognizing that

- (A1) $Y_{1,2} = Y_{1,2}(Z, X_{2,1}, Y_{2,1}, Y_{2,3})$
- (A2) $Y_{1,3} = Y_{1,3}(Z, X_3, Y_{3,1}, Y_{3,2})$
- (A3) $Y_{2,1} = Y_{2,1}(Z, X_1, Y_{1,2}, Y_{1,3})$
- (A4) $Y_{2,3} = Y_{2,3}(Z, X_3, Y_{3,1}, Y_{3,2})$

(A5)
$$Y_{3,1} = Y_{3,1}(Z, X_1, Y_{1,2}, Y_{1,3})$$

(A6) $Y_{3,2} = Y_{3,2}(Z, X_2, Y_{2,1}, Y_{2,3})$

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 chain-differentiation w.r.t. $x_{1,j}$ applied to eq. A3. It yields

A7)
$$D(Y_{2,1}, x_{1,j}) = d(Y_{2,1}, x_{1,j}) + d(Y_{2,1}, Y_1) D(Y_1, x_{1,j}) + d(Y_{2,1}, Y_3) D(Y_3, x_{1,j})$$

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) [A]
$$\{D(Y,z_k)\} = \{d(Y,z_k)\}$$

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 BB₂. 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,

(A9)
$$Y_{1,2} = Y_{1,2}(Z, X_{2,1}, Y_{2,1}, Y_{2,3})$$

(A10)
$$Y_{1,3} = Y_{1,3}(Z, X_3, Y_{3,1}, Y_{3,2})$$

(A11)
$$Y_{2,1} = Y_{2,1}(Z, X_1, Y_{1,2}, Y_{1,3})$$

(A12)
$$Y_{2,3} = Y_{2,3}(Z, X_3, Y_{3,1}, Y_{3,2})$$

(A13)
$$Y_{3,1} = Y_{3,1}(Z, X_1, Y_{1,2}, Y_{1,3})$$

(A14)
$$Y_{3,2} = Y_{3,2}(Z, X_2, Y_{2,1}, Y_{2,3})$$

(A15)
$$X_1 = X_1(Z, Y_{1,2}, Y_{1,3})$$

(A16) $X_2 = X_2(Z, Y_{2,1}, Y_{2,3})$

(A17)
$$X_3 = X_3(Z, Y_{3,1}, Y_{3,2})$$

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 $Y_{2,1}$ treated as a subset of Y_2 , we obtain

(A18)
$$D(Y_2,z_k) = d(Y_2,z_k) + d(Y_2,X_2)D(X_2,z_k) + d(Y_2,Y_1)D(Y_1,z_k) + d(Y_2,Y_3)D(Y_3,z_k)$$

and for X₂, again as one example:

(A19)
$$D(X_2, z_k) = d(X_2, z_k) + d(X_2, Y_1)D(Y_1, z_k) + d(X_2, Y_3)D(Y_3, z_k)$$

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 BB_r (including the option of finite differencing). The derivatives of X_r w.r.t. z_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 $D(Y_2, z_k)$ and $D(X_2, z_k)$ 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

$$(A20) [M_{yy}] \{D(Y,z_k)\} + [M_{yx}] \{D(X,z_k)\} = d(Y,z_k) [M_{xy}] \{D(Y,z_k)\} + [M_{xx}] \{D(X,z_k)\} = d(X,z_k)$$

The internal structure of the M-matrices in the above is

for $[M_{yy}]$:

Γι	$-d(Y_1, Y_2)$	$-\mathbf{d}(\mathbf{Y}_1,\mathbf{Y}_3)$
$\overline{-d(\mathbf{Y}_2,\mathbf{Y}_1)}$	I	$- d(Y_2, Y_3)$
$-d(Y_3, Y_1)$	$-d(Y_3, Y_2)$	Ι

for $[M_{yx}]$:

$-\mathbf{d}(\mathbf{Y}_1,\mathbf{X}_1)$	0	0]
0	$-d(Y_2, X_2)$	0
0	0	$-\mathrm{d}(\mathrm{Y}_3,\mathrm{X}_3)$

for $[M_{xy}]$:

0	$-d(X_1, Y_2)$	$- d(X_1, Y_3)$
$\overline{-d(X_2,Y_1)}$	0	$- d(X_2, Y_3)$
$\boxed{-\mathbf{d}(\mathbf{X}_3,\mathbf{Y}_1)}$	$-d(X_3, Y_2)$	0

and for $[M_{xx}]$:

I	0	0
0	I	0
0	0	Ι

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:

- M_{yy}, M_{yx}, and d(Y,z_k) ----- from BBSA
- M_{xy} and d(X,z_k) ----- from BBOSA

As in GSE, one may obtain $D(Y_2,z_k)$ and $D(X_2,z_k)$ for all z_k , k = 1--->NZ 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:	Р
	Find:	х
	Minimize:	(X,P)
	Satisfy:	$G(X,P) \le 0$, including side- constraints and move limits

where P are parameters kept constant while an optimizer manipulates X. z_k , and $Y_{r,s}$. because these quantities are kept constant in BBOPT_r.

After ϕ_{min} , and X_{opt} are found, one may seek sensitivity of these quantities to the change of P in form of the derivatives $D(\phi_{min}, P)$ and $D(X_{opt}, P)$.

Vanderplaats and Cai, 1986, review techniques, rigorous and approximate, available for calculating $D(X_{opt},P)$. The technique adapted for the BLISS/A purposes comprises the following steps executed for BB_{r} :

1. Choose parameter P_k , an element of Z or Y, and increment it by a ΔP

2. Use derivatives from SSA to extrapolate F and G_0

In the BLISS application, the parameters P in BB_r are

BB	Inputs	Internal	Outputs
Structures	AR, Λ , $\frac{V_{c}}{S_{REF}}$, W_{FO} , W_{O} , W_{E} , L, N_{Z} , λ , x	$t = \frac{t_c S_{REF}}{\sqrt{S_{REF}AR}}; b / 2 = \sqrt{S_{REF}AR / 2}; R = \frac{1 + 2\lambda}{3(1 + \lambda)}; \Theta =$ $pf(x, \frac{b}{2}, R, L); Fol = pf(x); W_w = (0.0051(W_T N_Z)^{0.557} S_{REF}^{0.649} + AR^{0.5}(\frac{b}{2})^{-0.4}(1 + \lambda)^{0.1}(0.1875S_{REF})^{0.1} / \cos(\Lambda))Fol; W_{Fw} =$ $(5S_{REF} / 18)(\frac{2}{3}t)42.5; W_F = W_{Fw} + W_{Fo}; W_T = W_o + W_w + W_F + W_E; \sigma 1 \to \sigma 5 = pf(\frac{t}{c}, L, x, \frac{b}{2}, R);$ Constraints	W _τ , W _F , Θ
A ero- dynamics	м, п, л,	$\begin{array}{c c} Constraints & \sigma 1 \to \sigma 5 \leq 1.09; 0.96 \leq \Theta \leq 1.04 \\ if h < 36089 ft, V = M1116.39 \sqrt{1 - (6.875e - 06)h}, \rho = (2.377e - 03)(1 - (6.875e - 06h))^{4.2561}; V = M968.1, \rho = (2.377e - 03)e^{-(h - 36089) 20806.7}, \\ if h > 36089 ft; C_{L} = \frac{W_{T}}{0.5\rho V^{2}S_{REF}}; Fo1 = pf(ESF, C_{f}); C_{Dmin} = \\ C_{Dmin, M < 1}Fo1 + 3.05(\frac{t}{C})^{-3}\cos(\Lambda)^{-2}; k = 1/(\pi 0.8AR); Fo2 = pf(\Theta); \\ C_{D} = (C_{Dmin} + kC_{L}^{-2})Fo2; L = W_{T}; D = C_{D}0.5\rho V^{2}S_{REF}; \\ dp / dx = pf(\frac{t}{C}) \\ \hline \end{array}$	L, D, <mark>L</mark> D
Propulsion	M, h, D, W _{BE} , T	$T = T * 16168.6; Temp = pf(M,h,T); ESF = (\frac{D}{3}) / \overline{T};$ SFC = 1.1324 + 1.5344M - (3.2956e - 05)h - (1.6379e - 04)T -031623M ² + (8.2138e - 06)Mh - (10.496e - 05)TM - (8.574e - 11)h ²	SFC, W _e , ESF
	$\begin{array}{c} M, h, \overset{L}{\underset{D}{\overset{D}{\overset{D}{\overset{T}}{\overset{T}{\overset{T}{\overset{T}{\overset{T}}{\overset{T}{\overset{T}{\overset{T}{\overset{T}{\overset{T}{\overset{T}{\overset{T}}{\overset{T}{\overset{T}}}}}}}}}$	$\theta = 1 - 6.875e - 06 * h, \text{ if } h < 36089ft; \ \theta = 0.7519$ if $h > 36089ft; R = \frac{M(L D)661\sqrt{\theta}}{SFC} ln\left(\frac{W_T}{W_T - W_F}\right)$	R
Constants Side Con- straints	$0.1 \le \lambda \le 0$	20001b; $W_0 = 250001b$; $N_z = 6g$; $W_{BE} = 43601b$; $C_{Dmin.M \le l} = 0.0$ 0.4; $0.75 \le x \le 1.25$; $0.75 \le C_f \le 1.25$; $0.1 \le T \le 1.0$; $0.01 \le \frac{1}{c} \le 0$ ≤ 60000 ; $1.4 \le M \le 1.8$; $2.5 \le AR \le 8.5$; $40 \le A \le 70$; $500 \le S_{Rl}$.09;

Table A1: BB Definitions

linearly and by Linear Programming solve

FindXMinimizeFSatisfy $G_o \le 0$ XL <= X <= XU; where XL and XU incorporate</td>the side constraints and the move limits;

to obtain X_{opt}.

- 3. Approximate $D(X,P) = \Delta X_{opt} / \Delta P$
- 4. Repeat from #1 for all elements of Z and Y input into BB_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 d(X,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:

(A22) $PF = A_0 + A_i^*S^T + (1/2)^*S^*A_{ii}^*S^T$

Where S is the vector of independent variables, and A_0 , A_i , and A_{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 A_{ij} are random numbers between 0 and 1. For this model, they are

		0.3970	0.8152	0.9230	0.1108
	0.4252		0.6357	0.7435	0.1138
A _{ij} =	0.0329	0.8856		0.3657	0.0019
	0.0878	0.7248	0.1978		0.0169
	0.8955	0.4568	0.8075	0.9239	

The remaining coefficient are:

- $\Theta \dots > A_o = [1.0]; A_i = [0.3 0.3 0.3];$ $A_{ii} = [0.4 - 0.4 - 0.4];$
- Fo1 ---> $A_o = [1.0]; A_i = [6.25]; A_{ii} = [0];$
- $\sigma_1 \longrightarrow A_o = [1.0]; A_i = [-0.75 \ 0.5 \ -0.75 \ 0.5]$

0.5]; $A_{ii} = [-2.5 \ 0 \ -2.5 \ 0 \ 0];$

- $\sigma^2 \longrightarrow A_o = [1.0]; A_i = [-0.5 \ 0.333 \ -0.5 \ 0.333]; A_{ii} = [-1.111 \ 0 \ -1.111 \ 0 \ 0];$
- $\sigma_{3--->}$ $A_o = [1.0]; A_i = [-0.375 \ 0.25 \ -0.375 \ 0.25 \ 0.25 \ 0.25]; A_{ii} = [-0.625 \ 0 \ -0.625 \ 0 \ 0];$
- $\sigma 4 \dots > A_{\sigma} = [1.0]; A_i = [-0.3 \ 0.2 \ -0.3 \ 0.2 \ 0.2];$ $A_{ii} = [-0.4 \ 0 \ -0.4 \ 0 \ 0];$
- $\sigma 5 \dots A_o = [1.0]; A_i = [-0.25 \ 0.1667 \ -0.25 \ 0.1667 \ 0.1667]; A_{ii} = [-0.2778 \ 0 \ -0.2778 \ 0 \ 0];$
- Fo2 ---> $A_o = [1.0]; A_i = [0.2 \ 0.2]; A_{ii} = [0 \ 0];$
- Fo3 ---> $A_o = [1.0]; A_i = [0]; A_{ii} = [0.04];$
- $dp/dx \rightarrow A_0 = [1.0]; A_i = [0.2]; A_{ii} = [0];$
- Temp ---> $A_o = [1.0]; A_i = [0.3 -0.3 0.3];$ $A_{ii} = [0.4 -0.4 0.4];$

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.

си		đ	tham I and anti-		:
ď				uiciii. Local upullilizationis are periorined on each DB (BBOP I) as well as a	well as a
5 6	The second to be a se	<u></u>	system level optim	system level optimization (SOP1) using a gradient guided path based on the	sed on the
s <i>P</i> o		8 8	Lagrange multiplie	Lagrange multipliers (OSAAA). Finally, all optimized changes to design	design
%	Name Page Number	° 6	var radics are used	variables are used to update the model for all milproved range.	
%		%	Author	: Jeremy S. Aote NASA I anoley/GWI1 6	Shring '98
%	BLISS. 16	%			or Sundo
%	SYSTEM_ANALYSIS19	%	Variables		
%	BB_WEIGHT	%	Α	- Coefficient matrix in GSE	none
%	BB_DRAGPOLAR23	%	DY AR	- Vector of total derivatives, behavior	
Ъ	BB_POWER	%	ł	variables wir t asnect ratio	Varv
%	BB_RANGE24	%	DY Cf	- Vector of total derivatives, behavior	ί my
%	POLYAPPROX25	%	I	variables w.r.t skin friction coefficient	varv
%	BB10PT26	%	DYE1_Z	- Matrix of total derivatives, behavior	
%	BBIWRAPPER28	%		variables from BB1 w.r.t Z variables	Varv
8	BB20PT28	%	DYE2_Z	- Matrix of total derivatives, behavior	•
6	Ē	%		variables from BB2 w.r.t Z variables	varv
<i>b</i> %	BB30PT	%	DYE3_Z	- Matrix of total derivatives, behavior	
8	BB3WRAPPER	%		variables from BB3 w.r.t Z variables	Varv
8	SYSOPT	%	DY4_Z	- Vector of total derivatives. range w.r.t	
8	SYSWRAPPER	%	I	Z variables	Varv
%	INBOUNDS	%	DY_h	- Vector of total derivatives, behavior	
8	FIN_DIFF	%		variables w.r.t altitude	Varv
6		%	DY_Lamda	- Vector of total derivatives, behavior	5
%	The electronic version of this code has been placed in custody of Dr. Jaroslaw	%		variables w.r.t wing sweep	vary
8 2	Sobieski, NASA Langley Research Center, Hampton, VA 23681.	%	DY_lamda	- Vector of total derivatives, behavior	•
8 8		%		variables w.r.t taper ratio	vary
0%		%	DY_M	- Vector of total derivatives, behavior	
		%		variables w.r.t Mach number	vary
ł		%	DY_Sref	- Vector of total derivatives, behavior	
 % 1		%		variables w.r.t wing surface area	vary
\$ 1		%	DY_T	- Vector of total derivatives, behavior	
% ?	Program BLISS	%		variables w.r.t throttle setting	vary
\$ 1	Ē	%	DY_tc	- Vector of total derivatives, behavior	I
\$ E	This program calls a system analysis for an aircraft range optimization	%		variables w.r.t thickness/chord ratio	vary
8 E	model, composed of the WEIGHI, DRAGPOLAR, and POWER black boxes	%	DY_x	- Vector of total derivatives, behavior	
8 8	(BBI, BB2, and BB3, respectively). Through black box (BBSA) and system	%		variables w.r.t wingbox x-section	vary
۶ t	sensitivity (33A) analyses, it calculates the derivatives necessary to solve the	%	DYX_nd	- Array of non-dimensional total derivatives,	
%	Global Sensitivity Equations (Sobieszczanski_Sobieski, 1990) and solves	%		behavior w.r.t. X variables	vary

<pre>% ************************************</pre>	%SSA% DY_lamda = A\dY_lambda; DY_x = A\dY_x; DY_Cf = A\dY_Cf; DY_T = A\dY_T;
s, vary l- be NM vary vary vary vary vary vary vary vary	X and Z ng Gauss- MATLAB
Array of non-dimensional total derivative behavior w.r.t. Z variables Difference between previous pass extrapo ated range and actual system analysis ran. Vector of total derivatives at the optimal state, range w.r.t Z variables Design variable initial values Change in range due to X variables Change in range due to X variables Vector of current design variables Change in range due to X and Z variables Vector of current design variables Upper bounds on design variables Wing taper ratio Upper bounds on design variables Wing taper ratio Wing taper ratio Wing taper ratio Wing taper ratio Wing taper ratio Wing surface area Altitude Mach number Aspect ratio Wing sweep Wing sweep Wing surface area Finds optimal change in X1 usin optimizer Finds optimal change in X3 usin optimizer Provides partial derivatives using	step forward finite differencing -Non-dimensionalizes bounds on -Solves for behavior variables usi Seidel iteration -Finds optimal change in Z using optimizer
DYZ_nd ext_error ext_error GRADphi4_Z i0 phi_BBOPT phi_SysOPT P_var phi_X_Z vlb vub X1(1) X1(1) X1(2)X	INbounds system_analysis Sys_OPT
	6 6 6 6 6 6

GRADphi4_Z = [Dphi1_Z + Dphi2_Z + Dphi3_Z + Dphi1_Y*DYE1_Z + Dphi2_Y*DYE2_Z + Dphi3_Y*DYE3_Z + DY4_Z] %SOPT%	[dZ,phi_SysOPT,G_sys(i.:)]=SysOPT(vlb_nd,vub_nd,i0,P_var,Y4,GRADphi4_Z,Z,dg 2_Z,G2);	phi_SysOPT=phi_SysOPT+Y4(1); ext_error = -phi_X_Z - Y4(1); phi_X_Z=-Y4(1)+phi_BBOPT+phi_SysOPT; %Non-dimensionalize Derivatives for Output Observation%	DYX = [DY_lamda DY_x DY_Cf DY_T]; DYZ = [DY_tc DY_h DY_M DY_AR DY_Lambda DY_Sref]; [XY_init,ZY_init] = NonDim(X1,X2,X3,Y1,Y2,Y3,Y4,Z); DYX_nd(:i) = DYX.*XY_init; DYZ_nd(:i) = DYZ.*ZY_init;	%Store Interim Results%	, Var(1:18,i)=[Y4 ext_error -phi_BB1OPT -phi_BB2OPT -phi_BB3OPT -phi_BBOPT -phi_SysOPT -phi_X_Z X1 X2 X3 Z];	%Update X and Z variables%	, X1=X1+dX1; X2=X2+dX2; X3=X3+dX3; 7=7+d7?	P_var=[X1 X2 X3 Z];	end %End BLISS loop	%Format Output Parameters%	RLB = Range_SSA ext_error dR_BB1 dR_BB2 dR_BB3 dR_X dR_Z Range_ext TapRat WingBox Cf Thrtl t/c h M AR lambda Sref';
DY_tc=A\dY_tc; DY_h=A\dY_h; DY_M=A\dY_M; DY_AP_A\dY_AP.	DY_Lambda=A\dY_Lambda; DY_Sref=A\dY_Sref;	DYE1_Z=[DY_tc(4) DY_h(4) DY_M(4) DY_AR(4) DY_Lambda(4) DY_Sref(4); DY_tc(8) DY_h(8) DY_M(8) DY_AR(8) DY_Lambda(8) DY_Sref(8)]; DYE2_Z=[DY_tc(1) DY_h(1) DY_M(1) DY_AR(1) DY_Lambda(1) DY_Sref(1); DY_tc(3) DY_h(3) DY_M(3) DY_AR(3) DY_Lambda(3) DY_Sref(3); DY_tc(9)	DY_h(9) DY_M(9) DY_AR(9) DY_Lambda(9) DY_Sref(9)]; DYE3_Z=[DY_tc(5) DY_h(5) DY_M(5) DY_AR(5) DY_Lambda(5) DY_Sref(5)]; DY4_Z=[DY_tc(10) DY_h(10) DY_M(10) DY_AR(10) DY_Lambda(10) DY_Sref(10)]; %BBOPT%	[vlb_nd,vub_nd]=INbounds(i0,vlb,vub);	[dX1,Lagrange1,phi_BB1OPT,BB1_G(i,:)]=BB1OPT(vlb_nd,vub_nd,i0,P_var,Z,Y21, Y31,X1,x_initial,L_initial,R_initial,Lift_initial,Twist_initial,tc_initial,C,DY_lamda, DY_x):	[dX2,Lagrange2,phi_BB2OPT]=BB2OPT(vlb_nd,vub_nd,i0,P_var,Z,Y12,Y32,X2,ES F initial Cf initial Twist initial to initial C DV C9.	[dX3,Lagrange3,phi_BB3OPT,BB3_G(i,:)]=BB3OPT(vlb_nd,vub_nd,i0,P_var,Z,Y23, X3,M_initial,h_initial,T_initial,C,DY_T); phi_BBOPT = phi_BB1OPT + phi_BB2OPT + phi_BB3OPT;	%0SAAA%	Dphi1_Z = [Lagrange1*dg1_Z]; Dphi2_Z = 11_agrange2*do7_Z1.	Dphi3_Z = [Lagrange3'*dg3_Z]; Dphi1_Y = [1_agrange1'*do1_VE1].	$Dphi2_Y = [Lagrange2*dg2_YE2];$ Dphi3_Y = [Lagrange3*dg3_YE3];

CLB = 'Pass_1 Pass_2 Pas Pass_11 Pass_12 Pass_13 Pass_20 Pass_21 Pass_22 Pass_29 Pass_30 Pass_40' Pass_38 Pass_40'	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_38 Pass_30 Pass_40; Pass_38 Pass_39 Pass_40;	s_9 Pass_10 ass_19 ass_37 ass_37	* * * * * * *	Z(2) Z(3) Z(4) Z(5) Z(6)	 Altitude Mach number Aspect ratio Wing sweep Wing surface area 	ft none deg ft ²
printman var, [], ALD, CLD);	9);			Output Variables		
RLBI = YI(1) YI(2) YI	RLB1 = Y1(1) Y1(2) Y1(3) Y2(1) Y2(2) Y2(3) Y3(1) Y3(2) Y3(3) Y4';			ں۔	- Vector of constants	vary
CLB1 = X1(1) X1(2) X2 X3';	X3:		%	Ga	- Vector of constraint values in BBa $(a = 1,2,3)$	vary
%printmat(DYX_nd(:,:,1	%printmat(DYX_nd(:.:,1),Non-Dimensional D(Y,X)',RLB1,CLB1);		%	Ya	- Vector of behavior variables output from	
			%		from BBa $(a = 1, 2, 3)$	vary
			%	Yab	- Vector of behavior variables output from	
RLB2 = "Y1(1) Y1(2) Y1	RLB2 = "Y1(1) Y1(2) Y1(3) Y2(1) Y2(2) Y2(3) Y3(1) Y3(2) Y3(3) Y4';		%		BBa, input to BBb (a & $b = 1,2,3$)	vary
CLB2 = Z1 Z2 Z3 Z4 Z5 Z6	5 26		%	Y4	- Objective function output from BB4	MN
%printmat(DYZ_nd(:,:,1	%printmat(DYZ_nd(:,:,1),Non_Dimensional D(Y,Z)',RLB2,CLB2);		%	var_inititial	- preserved values for polynomial construction	
			%		(var differs depending on particular poly.)	vary
G=[BB1_G G_sys(:,1) BB3_G];	B3_G];		%			
RLB3 = Pass 1 Pass 2 1	RLB3 = Pass_1 Pass_2 Pass_3 Pass_4 Pass_5 Pass_6';		% K	Local Variables		
CLB3 = 'sig1 sig2 sig3 sign + sig2 sig3 sign + si	ESF_u ESF_l t	emp Throttle';	%	Lu	- Test variable used in G-S iteration for	
printmat(G, Constraints		•	%		convergence of lift	lb
			%	Weu	- Test variable used in G-S iteration for	
			%		convergence of engine weight	lb
g/o			%	ESFu	- Test variable used in G-S iteration for	
%			%		convergence of engine scale factor	none
%	Subfunction SYSTEM ANALYSIS		%			
%	I		% Si	Subfunctions		
	This subfunction uses Gauss-Seidel iteration on the aircraft range optimization	optimization	%	BB_weight	-Calculates a/c structural weights	
	model to compute behavior variables, given a set of design variables.	s. Black boxes	%	BB_dragpolar	ar -Calculates aerodynamic values	
% WEIGHT, DRAGPC	WEIGHT, DRAGPOLAR, and POWER are called.		%	BB_power	-Calculates propulsion values	
%			%	BB_range	-Calculates system objective function	
Author	: Jeremy S. Agte NASA Langley/GWU	Spring 98	18° 18			
% % Innut Variables			0%			
	- Design variable initial values	varv	func-			
% XI(1)	- Wing taper ratio	none	tion[Y	1,Y2,Y3,Y4,Y12,Y	tion[Y1,Y2,Y3,Y4,Y12,Y14,Y21,Y23,Y24,Y31,Y32,Y34,G1,G2,G3,C,Twist_initial,	vist_initial,
	- Wingbox x-sectional area as poly. funct.	p.f.	x_init	ial, L_initial, R_initia	x_initial,L_initial,R_initial,ESF_initial,Cf_initial,Lift_initial,tc_initial,M_initial,h_in	_initial,h_in
% X2	- Skin friction coefficient as poly. funct.	p.f.	itial,T	_initial]=system_an	itial, T_initial]=system_analyis(Z,X1,X2,X3,i0)	
<i>%</i> X3	- Throttle setting	none				
% Z(1)	- Thickness/chord ratio	none	[Y1,Y	2,Y3,Y4,Y12,Y14,	[Y1,Y2,Y3,Y4,Y12,Y14,Y21,Y23,Y24,Y31,Y32,Y34,C]=Y_variables;	

%Preserve initial values for polynomial calculations $%$	%file	$\%$ file = 'in_out1.dat';		
Twist_initial=Y12(2); x_initial=i0(2);	%wn	le_var(z, Y 1, Y 2, Y	%write_var(z, Y 1, Y 2, Y 3, Y 4, Y 12, Y 14, Y 21, Y 23, Y 24, Y 31, Y 32, Y 34, X 1, X 2, X 3, C, file) ;	2,X3,C,file)
tc_initial=i0(5);				
L_initial=sqrt(i0(8)*i0(10))/2;	<u>%</u>			
$K_{101111111} = (1+2^{*}10(1))/(3^{*}(1+10(1)));$ FCF initial-V32(1).	%			
Cf initial=1.2(1),	%		Subfunction BB_WEIGHT	
Liff. initial-MO1/1).		•		
M initial=i0(7):	 % b	his subfunction ca	1 his subfunction calculates the weight of the aircraft by structure and adds them	adds them
$h_{initial=i0(6)}$; T_initial=i0(4);		orden a total air ute functions repre	to obtain a total aircraft weight. It calls the subfunction POLYAPPROX to com- pute functions represented by polynomials.	X to com-
%Execute Gauss Seidel iteration on system to find Y variables $%$	88	Author	: Jeremy S. Agte NASA Langley/GWU Sprin	Spring 98
		Input Variables		
Lu = Y2I(1) + 10;	%	U	- Vector of constants	VALV
Weu=Y31(1)+10;	%	L_initial	th	ft j
ESFu= Y32(1)+10;	%	Lift_initial		-qi
while ((abs(Lu-Y21(1))>(Y21(1)*.001)) (abs(Weu-Y31(1))>(Y31(1)*.001))	%	R_initial	- Initial location of lift as fraction of halfspan n	none
(aus(E5Fu-T32(1))>(Y32(1)*.001))) TV31(1).	%	tc_initial	o chord ratio	none
LUE I 2.1(1.); WV2.1(1).	%	Twist_initial		p.f.
weu=101(1); DCEv2071).	%	x_initial	-sectional thickness	p.f.
ESru= 1 32(1);	%	X1(1)		none
	%	X1(2)	gbox x-sectional area as poly. funct.	p.f.
70Call Black BOXes%	%	Y21		, qi
	%	Y31		lb
	%	Z(1)	ss/chord ratio	none
DD_weight(z, 1 21, 1 31, X1, X_initial, L_initial, K_initial, Lift_initial, Twist_initial, tc_ini	%	Z(2)	- Altitude ft	ft
	%	Z(3)	- Mach number	none
	%	Z(4)		none
[12,121,123,124,02]=BB_dragpolar(Z,Y12,Y32,X2,ESF_initial,Cf_initial,Twist_in	%	Z(5)	- Wing sweep	deg
	%	Z(6)	e area	ft²č
$[13,134,131,132,03]$ = Bb_power(Z, Y23,X3,M_initial,h_initial,T_initial,C);				
[14]=BB_range(L, Y14, Y24, Y34);		Output Variables		
	%	G1(1)		p.f.
With much iterative unrichic to an effect of the second se	%	G1(2)		p.f.
% while post-lictative variable to output nie%	%	G1(3)	- Stress on wing p.	p.f.

	- Stress on wing - Stress on wing	p.f. p.f.	$Y1(3) = PolyApprox(S_initial1,S1,flag1,bound1);$ Y12(2) = Y1(3);
% G1(6) % Y1(1)	- Wing twist as constraint - Total aircraft weight	p.f. Ib	%Polynomial function calculating wingbox X-sectional thickness $%$
	- Fuel weight	lb	
	- Wing twist	p.f.	S_initial2=[x_initial];
	- Total aircraft weight	lb	S2=[X1(2)];
	- Wing twist	p.f.	flag2=[1];
% Y14(1)	- Total aircraft weight	lh dl	bound2=[.008];
% Y14(2)	- Fuel weight	qI	Fo=PolyApprox(S_initial2,S2,flag2,bound2);
			$W_wing = Fo^*(.0051^*((Y21(1)^*C(3))^{557})^*(Z(6)^{649})^*(Z(4)^{57})^*(Z(1)^{4})^*((1+$
% Local Variables			$X1(1)^{A}.1)^{*}((cos(Z(5)^{*}pi/180))^{A}-1)^{*}((.1875^{*}Z(6))^{A}.1));$
% L	- Halfspan	Ĥ	
% R	- Wing aerodynamic center	none	$W_fuel_wing = (5*Z(6)/18)*(2/3*t)*(42.5);$
% t	- Wing thickness	ft	$Y1(2) = C(1) + W_fuel_wing;$
% W_wing	- Weight of the wing	lb	$Y1(1) = C(2) + W_wing + Y1(2) + Y31(1);$
% W_fuel_wing	ving - Wing aerodynamic center	lb	Y12(1) = Y1(1);
%			Y14(1) = Y1(1);
% Subfunctions			Y14(2) = Y1(2);
	PolyApprox -Forms polynomial functions for desir	desired variables	
<i>d</i> ₆			%THIS SECTION COMPUTES THE TOTAL WEIGHT OF A/C%
%o			%THIS SECTION COMPUTES CONSTRAINT POLYNOMIAL FUNCTIONS $%$
func-			
tion[Y1,Y12,Y14,G1]=BB_we itial,Twist initial,tc_initial,C)	tion[Y1,Y12,Y14,G1]=BB_weight(Z,Y21,Y31,X1,x_initial,L_initial,R_initial,Lift_in itial.Twist_initial,tc_initial,C)	nitial,Lift_in	S_initial3=[tc_initial,Lift_initial,x_initial,L_initial,K_initial,]; S3=[Z(1),Y21(1),X1(2),L,R];
	х		flag3 = [4,1,4,1,1];
%THIS SECT	%7HIS SECTION COMPUTES THE TOTAL WEIGHT OF A/C%	-%	bound3 = [.1,.1,.1,.1,.1,.1]; G1(1)=PolyApprox(S_initial3,S3,flag3,bound3); %wing stress
t = Z(1)*Z(6)/sqrt(t = Z(1)*Z(6)/sqrt(Z(6)*Z(4)); %wing thickness		
L=sqrt($Z(4)*Z(6)$)/2; %halfspan	/2; %halfspan		S_initial4=[tc_initial,Lift_initial,x_initial,L_initial,R_initial];
R=(1+2*X1(1))/(3	R=(1+2*X1(1))/(3*(1+X1(1))); %wing aerodynamic center location		S4=[Z(1), Y2(1), X1(2), L, R]; $f_{13,64} = [A + 1 + 11]$.
%Polynomial	%Polynomial function calculating wing twist%		bound4 = [.15,.15,.15,.15]; G1(2)=PolvAmrox(S_initial4.S4.flag4.bound4): %wing stress
S initial l=[x initi	S initial1=[x initial.L initial.R initial.Lift initial]:		
S1=[X1(2),L,R,Y21(1)];	1(1)];		S_initial5=[tc_initial,Lift_initial,x_initial,L_initial,R_initial];
flag I = [2,4,4,3];	35 35I.		S5=[Z(1), Y21(1), X1(2), L, R]; $f_{1305} = [4 \ 1 \ 4 \ 1 \ 11^{-1}];$
100000 - 100000	()		

bound5 = [.2,.2,.2,.2,.2]; G1(3)=PolyApprox(S_initial5,S5,flag5,bound5); %wing stress		cale factor ss/chord ratio	
S_initial6=[tc_initial,Lift_initial,x_initial,L_initial,R_initial];		- Altitude ft - Mach number none	
30=[∠(1), 1∠1(1), Δ1(∠), L, K]; flame = [4] 4] 1].	% Z(4)		
11450 - [7,1,1,1,1,1], boundf - [75 75 75 35].			
G1(4)=PolyApprox(S_initial6,S6,flag6,bound6); %wing stress	% Z(6)	- Wing surface area \mathfrak{h}^2	
	% Output Variables		
S_initial7=[tc_initial,Lift_initial,x_initial,L_initial,R_initial];	% G2	- Pressure gradient	
S7=[Z(1), Y21(1), X1(2), L, R];			
flag7 = [4, 1, 4, 1, 1];	% Y2(2)	- Drag lb	
bound/ = [.3,.3,.3,.3];		- Lift-to-drag ratio	
GI(3)=PolyApprox(S_initial7,S7,flag7,bound7); %wing stress		- Lift lb	
		- Drag lb	
GI(6)=YI(3); %wing twist	% Y24	- Lift-to-drag ratio	
%11413 SEC 11UN CUMPUTES CONSTRAINT POLYNOMIAL FUNCTIONS%	Loca		
		- Coefficient of lift none	
	% CD	- Coefficient of drag	
q_{b}	% CDmin	fficient	
%	% k		
% Subfunction BB_DRAGPOLAR	% rho		-13
%			,
	%		
	% Subfunctions		
	% PolyApprox	-Forms polynomial functions for desired variables	ables
% Author : Jeremy S. Agte NASA Langlev/GWU Spring 98	% %		
% Input Variables :	func-		
% C - Vector of constants vary	tion[Y2,Y21,Y23,Y24,G2	tion[Y2,Y21,Y23,Y24,G2]=BB dragpolar(Z,Y12,Y32,X2) ESF initial Cf initial Twis	Twis
Cf_initial - Initial coefficient of friction	t initial.tc initial.C)		CI M T (
% ESF_initial - Initial engine scale factor			
tc_initial - Initial thickness to chord ratio	%THIS SECTION CC	%THIS SECTION COMPUTES THE TOTAL DRAG OF THE A/C%	
Twist_initial			
X2 - Coefficient of friction	if Z(2)<36089		
Y12(1) - Total aircraft weight	V = Z(3)*(1116.39*sq)	V = Z(3)*(1116.39*sqrt(1-(6.875e-06*Z(2))));	
% Y12(2) - Wing twist p.f.	rho = (2.377e-03)*(1-(rho = (2.377e-03)*(1-(6.875e-06*Z(2)))^4.2561;	

else V = Z(3)*968.1; rho = $(2.377e-03)*(.2971)*exp(-(Z(2)-36089)/20806.7);$ end	%% S_ir S3=	%THIS SECTION S_initial3=[tc_initial]; S3=[7(1)]·	%THIS SECTION COMPUTES CONSTRAINT POLYNOMIALS% S_initial3=[tc_initial]; S3=[7/1\)-
$CL = Y12(1)/(.5*rho*(V^2)*Z(6)); \%Lift Coefficient$	flag	flag3=[1]; bound3=[.25];	
%Polynomial function modifying CDmin for ESF and friction coefficient $%$	G2(1)=PolyApprox(S_i	G2(1)=PolyApprox(S_initial3,S3,flag3,bound3); %adverse pressure gradient
S_initial1=[ESF_initial,Cf_initial]; S1=[Y32(1),X2(1)]; flag1 = [1,1]; bound1 = [75, 25].	~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~	THIS SECTION	%THIS SECTION COMPUTES CONSTRAINT POLYNOMIALS%
Fol = PolyApprox(S_initial1,S1,flag1,bound1); Fol = PolyApprox(S_initial1,S1,flag1,bound1); CDmin = C(5)*Fo1 + 3.05*(Z(1)^(5/3))*((cos(Z(5)*pi/180))^(3/2));	* * * *		Subfunction BB_POWER
if Z(3)>= 1 k=Z(4)*(Z(3)^2-1)*cos(Z(5)*pi/180)/(4*Z(4)*sqrt(Z(5)^2-1)-2); else k=1/(pi*0.8*Z(4)):	* * * *	This subfunction calculates scale factor. It calls the sub represented by polynomials.	This subfunction calculates fuel consumption and engine weight as well as engine scale factor. It calls the subfunction POLYAPPROX to compute functions represented by polynomials.
end	% %	Author	: Jeremy S. Agte NASA Langley/GWU Spring 98
%Polynomial function modifying CD for wing twist%	%	Input Variables	
• •	%	U	- Vector of constants vary
S_initial2=[Twist_initial];	%	h_initial	
S2=[Y12(2)];	%	M_initial	- Initial Mach number
flag2=[5];	%	T_initial	etting
bound2=[.25];	6	X3	tle setting
Fo2=PolyApprox(S_initial2,S2,flag2,bound2);	в° 8	Y23	
$CD = Fo2*(CDmin + k*(CL^2));$	8 6	Z(1) Z(2)	- I hickness/cnord ratio - Altitude
$Y2(2) = .5*rho*(V^2)*CD*Z(6);$	%	Z(3)	- Mach number
Y2(3) = CL/CD;	%	Z(4)	
Y2(1)=Y12(1);	%	Z(5)	- Wing sweep
Y23(1)=Y2(2);	° 8	Z(6)	- Wing surface area
124(1)=12(2), V1(1)-V2(1)-	s 9	Outnut Variables	
	8	G3(1)	- Engine scale factor constraint
%THIS SECTION COMPUTES THE TOTAL DRAG OF THE A/C%	%	G3(2)	- Engine temperature
	6 6 6	Output Variables G3(1) G3(2)	: - Engine scale factor constraint - Engine temperature

Y3(2) = C(4)*(Y3(3)^1.05)*3; Y31(1) = Y3(2); Y34(1) = Y3(1); Y32(1) = Y3(3); %THIS SECTION COMPUTES SFC, ESF, AND ENGINE WEIGHT%			% Subfunction BB_RANGE % Subfunction BB_RANGE % This subfunction calculates the system objective function, range, from the Breguet % This subfunction calculates the system objective function, range, from the Breguet % This subfunction calculates the system objective function, range, from the Breguet % This subfunction % This subfunction % Author : Jeremy S. Agte % Author : Jeremy S. Agte NASA Langley/GWU Spring 98 h % Y14(1) : Total aircraft weight Ib % Y14(2) - Fuel weight Ib
 Throttle setting constraint Specific fuel consumption Engine weight Engine scale factor Engine scale factor Specific fuel consumption 	 Non-dimensional throttle setting none Non-dimensional throttle setting in none Vector of constant coefficients for upper limit on throttle setting surface fit none Vector of constant coefficients for SFC none surface fit none Thrust required lb Upper limit on throttle setting none Forms polynomial functions for desired variables 		%Surface fit to engine deck (obtained using least squares approx) $%%$ Surface fit to engine deck (obtained using least squares approx) $%s=[1.13238425638512 1.53436586044561 -0.00003295564466 -0.00016378694115 -0.31623315541888 0.000000410691343 -0.00005248000590 -0.00000000085740.00000000190214 0.0000001059951];Y3(1)=s(1)+s(2)*Z(3)+s(3)*Z(2)+s(4)*Dim_Throttle+s(5)*Z(3)^2+2*Z(2)*S(3)+s(6)+2*Dim_Throttle*Z(3)*s(7)+s(8)*Z(2)^2+2*Dim_Throttle+s(5)*S(9)+s(10)*Dim_Throttle^2;Y3(3) = (Thrust/3)/Dim_Throttle;$
G3(3) Y3(1) Y3(2) Y3(3) Y3(3) Y3(3) Y32 Y34	 % Local Variables % Dim_Throttle N % %	function[Y3,Y34,Y31,Y32,G3]=F %THIS SECTION COMPUT Thrust = Y23(1); Dim Throttla - Y23(1)*16168 6.	$\mathcal{F}_{1} = \mathcal{F}_{2} $

% Y24 % Y34 Z(1) Z(2)	- Lift-to-drag ratio - Specific fuel consumption - Thickness/chord ratio - Altitude ft	none % 1/hr % none % ft	qualitative respons positive linear (fla, negative nonlinear	qualitative response to changes in other variables. Possible relationships are positive linear (flag = 1), negative linear (flag = 3), positive nonlinear (flag = negative nonlinear (flag = 4), and parabolic (flag = 5).	s arc flag = 2),
$\begin{array}{c} & \mathbf{Z}(\mathbf{z}) \\ & \mathbf{Z}(3) \\ & \mathbf{Z}(4) \end{array}$	urber atio	ne	Author	: Jeremy S. Agte NASA Langley/GWU Spring '98	80' gr
		deg %	Inpu		
% Z(6)	- Wing surface area	7 %	flag S	- Indicates functional relationship btwn var. no	none
% Output Variables		~ %			vary
	- Range N	% WN	S_bound	- Vector of bounds used to control slope of	
%	0	%		the polynomial function (narrow = high slope) none	none
% Local Variables		%	S_new	- Vector of current values of independent	
% Theta	- Temperature ratio	none %		variables	vary
% 01		%	Output Variables		
2		%		- Vector of coefficients (2 nd term) nc	none
function[Y4]=BB_range(Z,Y14,Y24,Y34)	te(Z,Y14,Y24,Y34)	%	Aij	- Matrix of coefficients (3 rd term) nc	none
	•	%		- Scalar coefficient (1 st term) nc	none
%THIS SECTION	THIS SECTION COMPUTES THE A/C RANGE (Breguet)%	%	FF	- Value of synthetic variable or modifier no	none
		%			
if Z(2)<36089		%	Local Variables		
theta=1-0.00006875*Z(2);	15*Z(2);	%	Α	- Solution matrix for polynomial fitting eqns. no	none
else		%	в	- Lower y-axis bound on polynomial	none
theta=.7519;		%	q	- Upper y-axis bound on polynomial	none
end		%	F_bound	- Bounds for dependent variable; RHS of	
		%			none
Y4(1) = ((Z(3)*Y24(1)))	Y4(1) = ((Z(3)*Y24(1))*661*sqrt(theta)/Y34(1))*log(Y14(1)/(Y14(1)-Y14(2)));	2))); %	Mtx_shifted		none
		% 0		- vector of Landon constants used to fin on-	none
		° 6	SI	ndent variable lower bound	none
		<i>%</i>		- Vector of current values of independent	
g ₀		<i>%</i>	I	variables normalized by initial values no	none
%		%	So	- Standard independent variable midpoint no	none
%	Subfunction POLYAPPROX	%	S_shifted	- Vector of normalized values of independent	
%		%		variables shifted to an area near the origin no	none
% This subfunction c	This subfunction calculates polynomial coefficients to characterize the behavior of correit contrastic variables and function modifiers. Move limits for each	chavior %	Su	- Standard independent variable upper bound no	none
	polynomial are selected based on knowledge of each variable or modifier's				

b=2*a; end %DETERMINE BOUNDS ON FF DEPENDING ON SLOPE-SHAPE %CALCULATE POLYNOMIAL COEFFICIENTS (S-ABOUT ORIGIN) So=0; Su=So+S_bound(i); Su=So+S_bound(i);	Mtx_shifted = [1 Si Si^2; 1 So So^2; 1 Su Su^2]; F_bound = [15*a; 1; 1+.5*b]; A = Mtx_shifted\F_bound; Ao = A(1); Ai(i) = A(2); Ai(i) = A(2); %CALCULATE POLYNOMIAL COEFFICIENTS	1 L UP OFF DIAGONALS OF A 0.3970 0.8152 0.9230 0.4415 0.6357 0.7435 0.8856 0.8390 0.3657	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	%CALCULATE POLYNOMIAL FF = A0 + Ai*(S_shifted') + (1/2)*(S_shifted)*(Aij)*(S_shifted');
<pre>function [FF, Ao, Ai, Aij] = PolyApprox(S, S_new, flag, S_bound) for i = 1:size(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(i)=0.75; end</pre>	Sshifted(i) = S_norm(i) - 1; %shift S vector near origin %DETERMINE BOUNDS ON FF DEPENDING ON SLOPE-SHAPE a=0.1; b=a; if flag(i)==5	%CALCULATE POLYNOMIAL COEFFICIENTS (S-ABOUT ORIGIN) So=0; So=0; Sl=So-S_bound(i); Mtx_shifted = [1 Sl Sl^2; 1 So So^2; 1 Su Su^2]; Mtx_shifted = [1 Sl Sl^2; 1; 1+(.5*b)^2]; A = Mtx_shiftedVF_bound; Ao = A(1);	AI(1) = A(2); Aij(i,i) = A(3); %CALCULATE POLYNOMIAL COEFFICIENTS else switch (flag(i)) case 0 S_shifted(i) = 0; case 3 a=-a;	b=a; case 2 b=2*a; case 4 a=-a;

			% gstore0	raint values at beginning	
	Subfunction BB10PT			of BB1 optimization p.f.	<u> </u>
			% Lagrangel	- Vector of Lagrange multipliers from BB1 at	
This subfunction se performing a local	This subfunction serves as a shell for the Matlab 'constr' optimization routine performing a local optimization on the WEIGHT module.	n routine	% %	optimum	Z
0			% Local Variables		
Author	: Jeremy S. Agte NASA Langley/GWU Sp	Spring '98	% options	- see Matlab 'constr' function	
			% vlb	- Non-dimensional lower bounds on BB1 design	
Input Variables			%	variables noi	none
່ ບໍ	- Vector of constants	vary	% vub	- Non-dimensional upper bounds on BB1 design	
DY_lamda	- Vector of total derivatives, behavior	vary	%	variables not	none
DY_x	- Vector of total derivatives, behavior		<i>‰</i> x0	- Vector of non-dimensional starting points	
	variables w.r.t wingbox x-section	vary	%	for BB1 optimization not	none
iO	- Design variable initial values	vary	%		
L_initial	- Initial halfspan length	fi	% Subfunctions		
Lift_initial	- Initial lift	lb	% BBIWRAPPER	ī	nts
P_var	- Vector of current design variable values	vary	%	for BB1 optimization	
R_initial	- Initial location of lift as fraction of halfspan	none	% constr	- Matlab optimization routine	
tc_initial	- Initial thickness to chord ratio	none	%		
Twist_initial	- Initial wing twist	p.f.	щ <u></u>		
vlb_nd	- Non-dimensional lower bounds on design				
	variables	none	func-		
vub_nd	- Non-dimensional upper bounds on design		tion[dX1,Lagrange1,fs	tion[dX1,Lagrange1,fstore,gstore0]=BB1OPT(vlb_nd,vub_nd,i0,P_var,Z,Y21,Y31,X1	,Y31,X1
	variables	none	,x_initial,L_initial,R_i	,x_initial,L_initial,R_initial,Liff_initial,Twist_initial,tc_initial,C,DY_lamda,DY_x)	DY_x)
x_initial	- Initial wingbox x-sectional thickness	p.f.			
X1(1)	- Wing taper ratio	none	vlb=[vlb_nd(1) vlb_nd(2)];	(2)];	
X1(2)	- Wingbox x-sectional area as poly. funct.	p.f.	<pre>vub=[vub_nd(1) vub_nd(2)];</pre>	d(2)];	
Y21	- Lift	lb	x0=[X1(1)/i0(1)-1,X1(2)/i0(2)-1];	2)/i0(2)-1];	
Y31	- Engine weight	ĄI	options(1)=1;		
Z(1)	- Thickness/chord ratio	none	options(2)=.0001;		
Z(2)	- Altitude	ft	options(3)=.0001;		
Z(3)	- Mach number	none	options(4)=.001;		
Z(4)	- Aspect ratio	none	options(14)=1000;		
Z(5)	- Wing sweep	deg	options(17)=.01;		
Z(6)	- Wing surface area	\mathbf{ft}^2			
Output Variables			[fstore0,gstore0,dX10] initial,Lift_initial,Twis	[fstore0,gstore0,dX10]=BB1WRAPPER(x0,i0,P_var,Z,Y21,Y31,x_initial,L_initial,R_ initial,Lift_initial,Twist_initial,tc_initial,C,DY_lamda,DY_x);	initial,R_
			1		

[x,o Y31 x);	ptions,Lagrange1] ,x_initial,L_initial	[x,options,Lagrange1]=constr('BB1WRAPPER',x0,options,vlb,vub,[],i0,P_Y31,x_initial,L_initial,R_initial,Lift_initial,Twist_initial,tc_initial,C,DY_x);	,i0,P_var,Z,Y21, ,,DY_lamda,DY_	% dX1 % f % g	 Vector of optimal changes in X1 variables Value of objective function for BB1 optim. Vector of local constraint values 	vary NM p.f.
ď				Loca		
2 %				% 31gma_uA % Twist_lA	 Upper allowable limit for stress constraints Lower allowable limit for twist constraint 	none none
6 6		Subfunction BB1WRAPPER		% Twist_uA	- Upper allowable limit for twist constraint	none
%	This subfunction c	This subfunction computes the objective function and the constraints for the	for the	% Subfunctions		
8 8	local optimization	local optimization on the WEIGHT module.			-Calculates a/c structural weights	
8 8	Author	: Jeremy S. Agte NASA Langley/GWU Spi	Spring '98	% %		
8	Input Variables			func-		
%	с	- Vector of constants	vary	tion[f,g,dX1]=BB1WRAP	tion[f,g,dX1]=BB1WRAPPER(x,i0.P var.Z,Y21,Y31,x initial.L initial.R initial.I if	initial Lift
81	DY_lamda	- Vector of total derivatives, behavior	vary	initial, Twist_initial, tc_initial, C, DY_lamda, DY_x)	tial,C,DY_lamda,DY_x)	111-61111111-
% 8	DY_x	- Vector of total derivatives, behavior				
% t	9	variables w.r.t wingbox x-section	vary	X1=[i0(1)*(1+x(1)),i0(2)*(1+x(2))];	(1+x(2))];	
% t	10 1	- Design variable initial values	vary			
\$ 8	L_initial	- Initial halfspan length	ft	[Y1,Y12,Y14,G1]=BB_we	[Y1,Y12,Y14,G1]=BB_weight(Z,Y21,Y31,X1,x_initial,L_initial,R_initial,Lift_initial	Lift_initial
%	Lift_initial	- Initial lift	q	,Twist_initial,tc_initial,C);		
6	P_var	- Vector of current design variable values	vary			
60	R_initial	- Initial location of lift as fraction of halfspan	none	Sigma_uA=1.05;		
%	tc_initial	- Initial thickness to chord ratio	none	Twist_uA=1.03;		
8	Twist_initial	- Initial wing twist	p.f.	Twist_IA=.97;		
8	X	- Vector of non-dimensional design variables				
%		for BB1 optimization	none	g(1)=G1(1)/Sigma_uA-1;		
6	x_initial	- Initial wingbox x-sectional thickness	p.f.	g(2)=G1(2)/Sigma_uA-1;		
8	Y21	- Lift	lb	g(3)=G1(3)/Sigma_uA-1;		
%	Y31	- Engine weight	lb	g(3)=G1(4)/Sigma_uA-1;		
8	Z(1)	- Thickness/chord ratio	none	$g(5)=G1(5)/Sigma_uA-1;$		
%	Z(2)	- Altitude	ft	g(6)=G1(6)/Twist_uA-1;		
8	Z(3)	- Mach number	none	g(7)=Twist_IA/G1(6)-1;		
%	Z(4)	- Aspect ratio	none			
8	Z(5)	- Wing sweep	deg	dX1=[X1(1)-P_var(1) X1(2)-P_var(2)];	?)-P_var(2)];	
в В В	Z(6)	- Wing surface area	$\mathbf{\hat{H}}^2$	f=-([DY_lamda(10),DY_x(10)]*dX1');	10)]*dX1');	
	Outnut Variables					
	Output variation					

<i>%</i>				%	
%				% Local Variables :	
%		Subfunction BB2OPT		options	see Matlab 'constr' function
%				vlb	- Non-dimensional lower bounds on BB2 design
%	This subfunction set	This subfunction serves as a shell for the Matlab 'constr' optimization	ation routine	% variables	ubles none
%	performing a local (performing a local optimization on the DRAGPOLAR module.		duv	- Non-dimensional upper bounds on BB2 design
%	> -			% variables	tbles none
%	Author	: Jeremy S. Agte NASA Langley/GWU Spi	Spring '98	- 0x	Vector of non-dimensional starting points
%					for BB2 optimization none
%	Input Variables			%	
%	C	- Vector of constants	vary	Subfuncti	
%	Cf_initial	- Initial coefficient of friction	p.f.	% BB2WRAPPER	- Contains objective function and constraints
%	DY_Cf	- Vector of total derivatives, behavior		%	for BB2 optimization
%		variables w.r.t skin friction coefficient	vary	% constr	- Matlab optimization routine
%	ESF_initial	- Initial engine scale factor	none	%	
%	i0	- Design variable initial values	vary	фо	
%	P_var	- Vector of current design variable values	vary		
%	tc_initial	- Initial thickness to chord ratio	none	func-	
%	Twist_initial	- Initial wing twist	p.f.	tion[dX2,Lagrange2,fstore]=BB2	tion[dX2,Lagrange2,fstore]=BB2OPT(vlb_nd,vub_nd,i0,P_var,Z,Y12,Y32,X2,ESF_in
8	vlb_nd	- Non-dimensional lower bounds on design		itial, Cf_initial, Twist_initial, tc_initial, C, DY_Cf)	iitial,C,DY_Cf)
%		variables	none		
%	vub_nd	- Non-dimensional upper bounds on design		vlb=[vlb_nd(3)];	
%		variables	none	vub=[vub_nd(3)];	
%	X 2	- Coefficient of friction	p.f.	x0=[X2(1)/i0(3)-1];	
%		- Total aircraft weight	lb	options(1)=1;	
%	Y12(2)	- Wing twist	p.f.	options(2)=.0001;	
%		- Engine scale factor	none	options(3)=.0001;	
%	Z(1)	- Thickness/chord ratio	none	options(4)=.001;	
%		- Altitude	ft	options(14)=1000;	
%	Z(3)	- Mach number	none	options(17)=.01;	
%	Z(4)	- Aspect ratio	none		
%	Z(5)	- Wing sweep	deg	[x,options,Lagrange2]=constr('B	[x,options,Lagrange2]=constr('BB2WRAPPER',x0,options,vlb,vub,[],i0,P_var,Z,Y12,
%	Z(6)	- Wing surface area	\mathbf{ft}^{2}	Y32,ESF_initial,Cf_initial,Twist_initial,tc_initial,C,DY_Cf);	t_initial,tc_initial,C,DY_Cf);
%					
%	Output Variables			[fstore,gstore,dX2]=BB2WRAPF	[fstore,gstore,dX2]=BB2WRAPPER(x,i0,P_var,Z,Y12,Y32,ESF_initial,Cf_initial,Twi
%	dX2	- Vector of optimal changes in X2 variables	vary	st_initial,tc_initial,C,DY_Cf);	
%		- Value of objective function for BB2 optim.	WN		
6	Lagrange2	- Vector of Lagrange multipliers from BB2 at			
%		optimum	ININI		

 % This subfunction cc % local optimization o % Author % Input Variables 			<i>%</i>	DD_ui agpoiar Calculates aerodynamic values	
	I rule subunction computes the objective function and the constraints for the local optimization on the DRAGPOLAR module.	for the	ж.		
			func-		
	: Jeremy S. Agte NASA Langley/GWU SI	Spring '98	tion[f,g,dX2]=BB2WRA	tion[f,g,dX2]=BB2WRAPPER(x,i0,P_var,Z,Y12,Y32,ESF_initial,Cf_initial,Twist_ini tial to initial C DV C5	initial, Twist_ini
				1)	
	- Vector of constants	varv	$X_{2}=[i_{0}(3)*(1+x(1))]$		
% Cf_initial	- Initial coefficient of friction	p.f.			
% DY_Cf	- Vector of total derivatives, behavior	4	[Y2,Y21,Y23,Y24,G2	[Y2,Y21,Y23,Y24,G2]=BB dragpolar(Z,Y12,Y32,X2,ESF initial Cf initial Twist in	initial Twist in
	variables w.r.t skin friction coefficient	vary	itial,tc initial,C);		
% ESF_initial	- Initial engine scale factor	none			
% j0	- Design variable initial values	vary	Pg uA=1.04:		
% P_var	- Vector of current design variable values	vary	g(1)=G2(1)/Pg uA-1:		
	- Initial thickness to chord ratio	none			
% Twist_initial	- Initial wing twist	p.f.			
% x	- Vector of non-dimensional design variables	4	dX2=[X2(1)-P var(3)]:		
ц,	for BB2 optimization	none	f=-([DY Cf(10)]*dX2):		
	- Total aircraft weight	lb			
	- Wing twist	p.f.			
	- Engine scale factor	none	γ_{0}		
	- Thickness/chord ratio	none	%		
% Z(2)	- Altitude	ft	%	Subfunction RB 3OPT	
	- Mach number	none	%		
	- Aspect ratio	none	% This subfunction s	This subfunction serves as a shell for the Matlab 'constr' ontimization routine	tion routine
	- Wing sweep	deg	% performing a local	performing a local optimization on the POWER module.	
% Z(6)	- Wing surface area	ft²	<i>2</i> %	•	
			% Author	: Jeremy S. Agte NASA Langley/GWU S	Spring '98
% Output Variables			%	,)) -
% dX2	- Vector of optimal changes in X2 variables	vary	% Input Variables	•••	
% f	- Value of objective function for BB2 optim.	MN		- Vector of constants	Varv
% g	- Vector of local constraint values	p.f.	% DY T	- Vector of total derivatives behavior	(m.
		1		variables w.r.t throttle setting	Varv
% Local Variables			% h initial	- Initial altitude	É e
% Pg_uA	- Upper allowable limit on pressure gradient		% i0_	- Design variable initial values	varv
%	constraint	none	% M initial	- Initial Mach number	- uou

1		ector of current design variable values itial throttle setting on-dimensional lower bounds on design ariables on dimensional upper bounds on design ariables hrottle setting inckness/chord ratio hickness/chord ratio lititude fach number spect ratio fing surface area fing surface area fing surface area fing surface area fing surface area fing surface area fing surface objective function for BB3 optim. ector of local constraint values at beginning f BB3 optimization ector of local constraint values at beginning f BB3 optimization fon-dimensional lower bounds on BB3 design ariables fon-dimensional lower bounds on BB3 design ariables for BB3 optimization or BB3 optimization or BB3 optimization	P_var T_initial vlb_nd vub_nd vub_nd X3 Y23 Z(1) Z(2) Z(3) Z(3) Z(4) Z(3) Z(4) Z(3) Z(4) Z(3) Z(4) Z(3) Z(4) Z(4) Z(3) Z(4) Z(3) Z(4) Z(3) Z(4) Z(5) Z(6) Z(6) Z(6) Z(6) Z(6) Z(6) Z(6) Z(6
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 Initial throttle setting Non-dimensional lower bounds on design variables Non-dimensional upper bounds on design Non-dimensional upper bounds on design none Throttle setting Throttle setting Drag Thickness/chord ratio Altitude Altitude Aspect ratio none 	options($ 1\rangle = .01$;		Z(5)
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 Initial throttle setting Non-dimensional lower bounds on design Non-dimensional upper bounds on design Non-dimensional upper bounds on design Non-dimensional upper bounds on design Introttle setting Throttle setting Drag Thickness/chord ratio Altitude Mach number 	options(14)=1000;		Z(4)
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 Initial throttle setting Non-dimensional lower bounds on design Non-dimensional upper bounds Non-dimensional upper bounds 	options(4)=.001;		Z(3)
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 Initial throttle setting Non-dimensional lower bounds on design Non-dimensional upper bounds on design Non-dimensional upper bounds on design Nonectional Throttle setting Drag 			Z (1)
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 Initial throttle setting Non-dimensional lower bounds on design Non-dimensional upper bounds on design Non-dimensional upper bounds on design Throttle setting 	ontions(1)=1.		V73
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 Initial throttle setting Non-dimensional lower bounds on design variables Non-dimensional upper bounds on design variables 			0.11
 Initial throttle setting Non-dimensional lower bounds on design variables Non-dimensional upper bounds on design 	vub=[vub_nd(4)];		
 Initial throttle setting Non-dimensional lower bounds on design variables Non-dimensional upper bounds on design 			
 Initial throttle setting Non-dimensional lower bounds on design variables 	vlb=[vlb_nd(4)];	- Non-dimensional upper bounds on design	vub nd
ial - Initial throttle setting - Non-dimensional lower bounds on design variables			
ial - Initial throttle setting none - Non-dimensional lower bounds on design			
ial - Initial throttle setting none - Non-dimensional lower buinds on design			
ial - Initial throttle setting none	nitial,h initial,T initial,C,DY_T)	- Non-dimensional lower bounds on design	vlh nd
international contract contract of the contrac			I_IIIIUAI
	tion[dX3.Lagrange3.fstore.gstore0]=BB3OPT(vlb nd.vub nd.i0.P var.Z,Y23,X3,M_i		T initial
- Vector of current design variable values val		gli variauje values	r_var
	-		-

	Subfunction SYSOPT	This subfunction serves as a shell for the Matlab 'constr' optimization routine performing a system optimization. Author : Jeremy S. Agte NASA Langley/GWU Spring '98	: - Derivative of BB2 constraint wrt Z p.f. - Pressure gradient p.f. - Vector of total derivatives at the optimal	e w.r.t Z variables riable initial values current design variable values nsional lower bounds on design	variables none - Non-dimensional upper bounds on design none variables none - Range NM - Thickness/chord ratio none	e arca
	Subfunct	serves as a shell for em optimization. : Jeremy S. Agte	•• • • •	state, ra - Design - Vector - Non-dii	variables - Non-dime variables - Range - Thickness	 Altitude Mach number Aspect ratio Wing sweep Wing surface :
Temp_uA=1.02; g(1)=G3(1)/ESF_uA-1; g(2)=ESF_1A/G3(1)-1; g(3)=G3(2)/Temp_uA-1; g(4)=G3(3); dX3=[X3(1)-P_var(4)]; f=-([DY_T(10)]*dX3);		This subfunction serves as a shell performing a system optimization. Author : Jeremy S. A.	Input Variables dg2_Z G2 GRADphi4_Z	i0 P_var vlb_nd	vub_nd Y4 Z(1)	Z(2) Z(3) Z(4) Z(5) Z(6) Output Variables
Ter g (3) g (3) g (3) g (2) g (3) g (2) g (3) g (3	2 8 8 8	~ & & & & & &	* * * * * * *	88888		& & & & & & & & & & & & & & & & & & & &
vary none vary none lb ff none none	deg ff ²	vary NM p.f.	none none none		nitial,C,DY	al,C);
 Design variable initial values Initial Mach number Vector of current design variable values Initial throttle setting Vector of non-dimensional design variables for BB3 optimization Drag Thickness/chord ratio Altitude Mach number Aspect ratio 	- Wing sweep - Wing surface area	: - Vector of optimal changes in X3 variables - Value of objective function for BB3 optim. - Vector of local constraint values	: - Lower allowable limit for ESF constraint - Upper allowable limit for ESF constraint - Upper allowable limit for temp. constraint	: -Calculates propulsion values	func- tion[f,g,dX3]=BB3WRAPPER(x,i0,P_var,Z,Y23,M_initial,h_initial,T_initial,C,DY T)	X3=[i0(4)*(1+x(1))]; [Y3,Y34,Y31,Y32,G3]=BB_power(Z,Y23,X3,M_initial,h_initial,T_initial,C); ESF_uA=1.5; ESF_IA=.5;
		bles	A les	ower	WRA	G3]=(
	% Z(5) % Z(6) %	 % Output Variables % dX3 % f % g 	 Cocal Variables Local Variables ESF_IA ESF_UA Temp_uA 	 % Subfunctions % BB_power %	func- tion[f,g,dX3]=BB3 T)	X3=[i0(4)*(1+x(1))]; [Y3,Y34,Y31,Y32,G3 ESF_uA=1.5; ESF_IA=.5;

885	dZ fstore	 Vector of optimal changes in Z variables Value of objective function for system optim. 	vary NM	[fstore,gstore,dZ]=9	ysWRAPPER(x,i0	[fstore,gstore,dZ]=SysWRAPPER(x,i0,P_var,Y4,GRADphi4_Z,dg2_Z,G2);	G2);
88	gstoreU	- vector or constraint values at beginning of system optimization	vary				
	T			9/0 01,			
	Local variables	······································		2	Cubbinetion	Cithfunction CVCW/D ADDFD	
8 8	options	- see Mattao Constratuticuon Mon-dimensional louer bounds on 7	-	8	SUDIUIU		
5 6	VID		euou		n committee the ob-	This subfine tion commutes the objective function and constraints for the	or the
8 6	qui	Vallautes - Non-dimensional unner bounds on 7			ation.		
5 6	100	- Ivon-unitoriatorial appen ocanate or o	enon				
8 6	×0	- Vector of non-dimensional starting points		% Author	: Jeremy S. Agte	NASA Langley/GWU	Spring '98
5 6		for system ontimization	none		7	`))
5 6				% Input Variables			
5	Subfunctions			% dg2_Z		- Derivative of BB2 constraint wrt Z	p.f.
6	SvsWRAPPER	ER - Contains objective function and constraints		% G2	- Pressure gradient	radient	p.f.
%	2			% GRADphi4_Z	'	Vector of total derivatives at the optimal	
%	constr	- Matlab optimization routine		%	state, rang	state, range w.r.t Z variables	vary
60				% i0	- Design va	Design variable initial values	vary
g					- Vector of	Vector of current design variable values	vary
				% x	- Vector of	Vector of non-dimensional design variables	s
func-	<u>,</u>			%	for systen	for system optimization	none
tion	{dZ,fstore,gstore0]=S	tion {dZ,fstore, gstore0]=SysOPT(vlb_nd,vub_nd,i0,P_var,Y4,GRADphi4_Z,Z,dg2_Z,	Z,Z,dg2_Z,	% Y4	- Range		MN
G2)	, ,		,	% Z(1)	- Thickness	Thickness/chord ratio	none
Ì				% Z(2)	- Altitude		Ĥ
-dly	vlb=[vlb_nd(5:10)];			% Z(3)	- Mach number	nber	none
vub	vub=[vub_nd(5:10)];			% Z(4)	- Aspect ratio	tio	none
=0x	[Z(1)/i0(5)-1,Z(2)/i0(x0=[Z(1)/i0(5)-1,Z(2)/i0(6)-1,Z(3)/i0(7)-1,Z(4)/i0(8)-1,Z(5)/i0(9)-1,Z(6)/i0(10)-1];	(0(10)-1];		- Wing sweep	ep	deg
opti	options(1)=1;				- Wing surface area	face area	ft
opti	options(2)=.0001;						
opti	options(3)=.0001;			Out	••		
opti	options(4)=.001:			% dZ	- Vector of	Vector of optimal changes in Z variables	vary
opti	options(14)=1000;			% f	- Value of c	Value of objective function for system optim.	m. NM
opti	options $(17)=.01$;			% g	- Vector of	Vector of constraint values	vary
•				%			
[fsto	pre0,gstore0,dZ0]=Sy	[fstore0,gstore0,dZ0]=SysWRAPPER(x0,i0,P_var,Y4,GRADphi4_Z,dg2_Z,G2);	Z,G2);	% Local Variables			
	I				- Used to c	- Used to construct move limits	none
ž ž	=constr('SysWRAPPE	[x]=constr('SysWRAPPER',x0,options,vlb,vub,[],i0,P_var,Y4,GRADphi4	Dphi4_Z,dg2_Z,G	% Pg_uA %	- Upper allo constraint	 Upper allowable limit on pressure gradient constraint 	none
()				2			2 2 2

%		8° 8	vub	- Non-dimension	Non-dimensional upper bounds on BB1 design	
function[f,g,dZ]=SysWRAPPER(x,i0,P_var,Y4,GRADphi4_Z,dg2_Z,	/ar,Y4,GRADphi4_Z,dg2_Z,G2)	5 8 8 8	x0	valators - Vector of non-dimens for BB1 optimization	variatores Vector of non-dimensional starting points for BB1 optimization	s none none
Z = [i0(5)*(1+x(1)),i0(6)*(1+x(2)),i0(7)*(1+x(3)),i0(8)*(1+x(4)), i0(9)*(1+x(5)),i0(10)*(1+x(6))];	<pre>*(1+x(3)),i0(8)*(1+x(4)),</pre>		Output Variables vlb_nd	: - Non-dimension	: - Non-dimensional lower bounds on design	_
dZ = [Z(1)-P_var(5) Z(2)-P_var(6) Z(3)-1 P_var(10)];	dZ = [Z(1)-P_var(5) Z(2)-P_var(6) Z(3)-P_var(7) Z(4)-P_var(8) Z(5)-P_var(9) Z(6)- P_var(10)];	8883	pu_duv	variables - Non-dimension variables	variables - Non-dimensional upper bounds on design variables	none
a=.2; Pg_uA=1.04; G2(1)=G2(1)/Pg_uA-1;		% % function	[vlb_nd,vub_nc	% %	vub)	
$g(1)=G2(1) + dg2_Z(1,1)*dZ(1); g(2)=abs(dZ(1))/(a*i0(5))-1; g(3)=abs(dZ(2))/(a*i0(6))-1; g(4)=abs(dZ(3))/(a*i0(6))-1; g(5)=abs(dZ(4))/(a*i0(9))-1; g(6)=abs(dZ(5))/(a*i0(9))-1; g(7)=abs(dZ(6))/(a*i0(10))-1; g(7)=abs$		vlb_nd= 1,vlb(6) vub_nd= 1,vub(6)	[vlb(1)/x0(1)-1. /x0(6)-1,vlb(7)/ =[vub(1)/x0(1)- //x0(6)-1,vub(7)	vlb(2)/x0(2)-1,vlb(3) k0(7)-1,vlb(8)/x0(8)- (,vub(2)/x0(2)-1,vub /x0(7)-1,vub(8)/x0(8	vlb_nd=[vlb(1)/x0(1)-1,vlb(2)/x0(2)-1,vlb(3)/x0(3)-1,vlb(4)/x0(4)-1,vlb(5)/x0(5)- 1,vlb(6)/x0(6)-1,vlb(7)/x0(7)-1,vlb(8)/x0(8)-1,vlb(9)/x0(9)-1,vlb(10)/x0(10)-1]; vub_nd=[vub(1)/x0(1)-1,vub(2)/x0(2)-1,vub(3)/x0(3)-1,vub(4)/x0(4)-1,vub(5)/x0(5)- 1,vub(6)/x0(6)-1,vub(7)/x0(7)-1,vub(8)/x0(8)-1,vub(9)/x0(9)-1,vub(10)/x0(10)-1]; %	lb(5)/x0(5)- x0(10)-1]; 1,vub(5)/x0(5)-))/x0(10)-1];
f = -(Y4(1) + GRADphi4_Z*dZ);			s subfunction ca erencing to calc	Subfunct Ils several subfuncti ulate the derivatives	Subfunction FIN_DIFF This subfunction calls several subfunctions that use one-step forward finite differencing to calculate the derivatives required by the BLJSS method.	ard finite ethod.
% Subfunction INBOUNDS	BOUNDS		Author	: Jeremy S. Agte	NASA Langley/GWU	Spring '98
 % This subfunction calculates the non-dimensional upper and lower % design variables. 	limensional upper and lower bounds of the	% Input: % C	ıt: C Cf_initial	- Vector of constants - Initial coefficient of friction	nts It of friction	vary p.f.
% Author : Jeremy S. Agte	NASA Langley/GWU Spring '98	6 6 6	ESF_initial G1(1)	 Initial engine scale factor Stress on wing 	ale factor	none p.f.
 % Input Variables : % vlb - Non-dimensic % 	: - Non-dimensional lower bounds on BB1 design variables none		G1(3) G1(4) G1(5)	- Suress on wing - Stress on wing - Stress on wing - Stress on wing		р.г. р.f. р.f.

ft none deg ft ²	none	vary	vary	vary	vary	vary	vary	vary	vary	vary		vary	vary	vary	Varv		vary		vary
 Altitude Mach number Aspect ratio Wing sweep Wing surface area 	- Coefficient matrix in GSE - 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	w.r.t Y variables entering BB1 - Vector of derivatives, BB2 constaints	w.r.t Y variables entering BB2 - 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, benavior variables w.r.t skin friction coefficient	 Vector of partial derivatives, behavior 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 nartial 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
Z(2) Z(3) Z(4) Z(5) Z(6) Outrout:	A del Z	de2 Z	dg3_Z	dg1_YE1	dg2_YE2	dg3 YE3	dY AR		dY_CT	dY_h	dY_Lamda	dY_lamda	MAP		dY_Sref	dY_T		dY_tc	dY_x
• * * * * * * * *		2 8 8	8 8	8 8	r 8 b	8 8	r 6 r	6 2	8 8	8 8	%	6 6	8 8	2 6	Ъ В В	2 b	%	%	%
p.f. p.f. p.f. ft ft	lb none	none	none p.f.	p.f. Ib	lb p.f.	lb D.f.	.e.e	୍ୟ :	lb none	ଶ ଶ	none	1/hr Ib	none	none	1/hr NM	none	p.f.	p.f.	none
 Wing twist as constraint Pressure gradient Engine scale factor constraint Engine temperature Throttle setting constraint Initial altitude Initial halfsnan length 	- Initial lift - Initial Mach number	 Initial Adact number Initial location of lift as fraction of halfspan Initial thickness to chord ratio 	- Initial throttle setting - Initial wing twist	 Initial wingbox x-sectional thickness Total aircraft weight 	- Fuel weight - Wing twist	- Total aircraft weight - Wing twist	- Total aircraft weight - Finel weight	- Lift	- Drag - Lift-to-drag ratio	- Lift - Drag	- Lift-to-drag ratio	 Specific fuel consumption Engine weight 	- Engine scale factor	- Engine weight - Engine scale factor	- Specific fuel consumption	- Objective function output from the	- Wingbox x-sectional area as poly. funct.	- Skin friction coefficient as poly. funct.	 Throttle setting Thickness/chord ratio
G1(6) G2 G3(1) G3(2) G3(3) h_initial	Lift_initial M initial	R_initial rc initial	Tinitial Twist_initial	x_initial Y1(1)	Y1(2) Y1(3)	Y12(1)	$Y_{14(2)}$	Y2(1)	Y2(2) Y2(3)	Y21 Y23	Y24	Y3(1) Y3(2)	Y3(3)	Y32	Y34 V4	14 X1(1)	X1(2)	X 2	X3 Z(1)
6 8 8 8 8 8 6	2 6 6	8 8 B	2 B B	:	в В В	2 B B	2 6 6	e 19	° 8	%	26	% %	8° 8	° %	ъ В	% %	%	%	°° %

		f
Subfunctions :		tunc- tion[A.dY lambda.dY x.dY Cf.dY T.dY tc.dY h.dY M.dY AR.dV I ambda.dV S
fin_diff_A12	-Calculates the A12 submatrix of GSE eqns.	ref.dg1_Z,dg2_Z,dg3_Z,dg1_YE1,dg2_YE2,dg3_YE31=FIN_DIFF(Z_Y1_Y2_Y3_Y4
fin_diff_A13	-Calculates the A13 submatrix of GSE equs.	Y12,Y14,Y21,Y23,Y24,Y31,Y32,Y34,X1,X2,X3,G1 G7 G3 C Twist initial v initial
fin_diff_A21	-Calculates the A21 submatrix of GSE equs.	L initial.R initial.ESF initial.Cf initial Lift initial to initial M initial h initial T in
fin_diff_A23	-Calculates the A23 submatrix of GSE equs.	
fin_diff_A32	-Calculates the A32 submatrix of GSE eqns.	
fin_diff_A41	-Calculates the A41 submatrix of GSE eqns.	%calculate Y partials%
fin_diff_A42	-Calculates the A42 submatrix of GSE eqns.	
fin_diff_A43	-Calculates the A43 submatrix of GSE eqns.	[A12]=fin_diff_A12(Z,Y1,Y21,Y31,X1,C.x_initial.L_initial.R_initial.I_iff_initial_Twi
fdG1_Y21	-Calculates BB1 constraints w.r.t Y variables	st_initial,tc_initial);
	coming into BB1 from BB2; derivative	[A13]=fin_diff_A13(Z,Y1,Y21,Y31,X1,C,x_initial_L_initial_R_initial_I iff_initial_Twi
fdG1_Y31	-Calculates BB1 constraints w.r.t Y variables	st_initial,tc_initial);
	coming into BB1 from BB3; derivative	[A21]=fin_diff_A21(Z,Y2,Y12,Y32,X2,C,Twist_initial.ESF_initial.Cf_initial.tc_initi
fdG2_Y12	-Calculates BB2 constraints w.r.t Y variables	al);
	coming into BB2 from BB1; derivative	[A23]=fin_diff_A23(Z_Y2_Y12_Y32_X2_C_Twist_initial ESF_initial Cf_initial te_initi
fdG2_Y32	-Calculates BB2 constraints w.r.t Y variables	
	coming into BB2 from BB3; derivative	[A32]=fin_diff_A32(Z,Y3,Y23,X3,C,M_initial,h_initial,T_initial):
fdG3_Y23	-Calculates BB3 constraints w.r.t Y variables	[A41]=fin_diff_A41(Z,Y4,Y14,Y24,Y34);
	coming into BB3 from BB2	[A42]=fin_diff_A42(Z,Y4,Y14,Y24,Y34);
fdY1_X1	-Calculates Y1 output w.r.t. change in X variables	[A43]=fin_diff_A43(Z,Y4,Y14,Y24,Y34);
fdY2_X2	-Calculates Y2 output w.r.t. change in X variables	[dg1_Y21]=fdG1_Y21(Z,Y1,Y21,Y31,X1,G1,C,x_initial,L_initial,R_initial,Lift_initi
	tor BB2	al,Twist_initial,tc_initial);
fdY3_X3	-Calculates Y3 output w.r.t. change in X variables	[dg1_Y31]=fdG1_Y31(Z,Y1,Y21,Y31,X1,G1,C,x_initial,L_initial,R_initial,Lift_initi
	for BB3	al,Twist_initial,tc_initial);
td Y L_Z	-Calculates Y1 output w.r.t. change in Z variables	dg1_YE1 = [dg1_Y21 dg1_Y31];
fdY2_Z	-Calculates Y2 output w.r.t. change in Z variables	[dg2_Y12]=fdG2_Y12(Z,Y2,Y12,Y32,X2,G2,C,Twist_initial,ESF_initial,Cf_initial,t
fdY3_Z	-Calculates Y3 output w.r.t. change in Z variables	c_initial);
fdY4_Z	-Calculates Y4 output w.r.t. change in Z variables	[dg2_Y32]=fdG2_Y32(Z,Y2,Y12,Y32,X2,G2,C,Twist_initial FSF_initial Cf_initial t
fdG1_Z	-Calculates BB1 contraint output w.r.t. change in	c_initial);
	Z variables	dg2 YE2 = [dg2 Y12 dg2 Y32]:
fdG2_Z	-Calculates BB2 contraint output w.r.t. change in	[dg3 Y23]=fdG3 Y23/Z Y3 Y3 Y3 G3 C M initial h initial T initial)
		de3 YE3 = $[de3$ Y23]:
fdG3_Z	-Calculates BB3 contraint output w.r.t. change in	
		%construct A matrix%

 $\begin{array}{c} 1 \ 0 \ 0 \ - Al2(1,1) \ 0 \ 0 \ 0 \ - Al3(1,1) \ 0 \ 0 \\ 0 \ 1 \ 0 \ - Al2(2,1) \ 0 \ 0 \ 0 \ - Al3(2,1) \ 0 \ 0 \\ 0 \ 0 \ 1 \ - Al2(2,1) \ 0 \ 0 \ 0 \ - Al3(3,1) \ 0 \ 0 \\ - A2l(1,1) \ 0 \ - A2l(1,2) \ 1 \ 0 \ 0 \ 0 \ 0 \ - A23(1,1) \ 0 \\ - A2l(2,1) \ 0 \ - A2l(2,2) \ 0 \ 1 \ 0 \ 0 \ 0 \ - A23(1,1) \ 0 \\ - A2l(3,1) \ 0 \ - A2l(3,2) \ 0 \ 1 \ 0 \ 0 \ - A23(3,1) \ 0 \\ - A2l(3,1) \ 0 \ - A2l(3,2) \ 0 \ 0 \ 0 \ 0 \ - A23(3,1) \ 0 \\ 0 \ 0 \ 0 \ 0 \ - A22(1,1) \ 0 \ - A22(3,1) \ 0 \\ 0 \ 0 \ 0 \ 0 \ - A23(3,1) \ 0 \\ - A2l(3,1) \ 0 \ - A22(3,1) \ 0 \ 0 \ 0 \ 0 \ - A23(3,1) \ 0 \\ - A2l(3,1) \ 0 \ - A2l(3,2) \ 0 \ 0 \ 0 \ 0 \ - A23(3,1) \ 0 \\ - A2l(1,1) \ - A4l(1,2) \ 0 \ 0 \ 0 \ 0 \ - A42(1,1) \ - A43(1,1) \ 0 \ 0 \ 1];$

%-----calculate X partials-----%

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

%-----calculate Z partials-----%

[dY1_Z1,dY1_Z4,dY1_Z5,dY1_Z6]=fdY1_Z(Z,Y1,Y21,Y31,X1,C,x_initial,L_initial, R_initial,Lift_initial,Twist_initial,tc_initial); [dY2_Z1,dY2_Z2,dY2_Z3,dY2_Z4,dY2_Z5,dY2_Z6]=fdY2_Z(Z,Y2,Y12,Y32,X2,C, Twist_initial,ESF_initial,Cf_initial,tc_initial); [dY3_Z2,dY3_Z3]=fdY3_Z(Z,Y3,Y23,X3,C,M_initial,h_initial,T_initial); [dY4_Z2,dY4_Z3]=fdY4_Z(Z,Y4,Y14,Y24,Y34); [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 aby

[dg3_Z]=fdG3_Z(Z,Y3,Y23,X3,G3,C,M_initial,h_initial,T_initial);

%-----construct RHS matrices-----%

dY_lambda = [dY1_X1_1; 0; 0; 0; 0; 0; 0; 0]; dY_x = [dY1_X1_2; 0; 0; 0; 0; 0; 0; 0]; dY_Cf = [0; 0; 0; dY2_X2; 0; 0; 0; 0];

 $dY_T = [0; 0; 0; 0; 0; 0; dY3_X3; 0];$

dY_tc = [dY1_Z1' dY2_Z1' 0 0 0]; dY_h = [0 0 0 dY2_Z2' dY3_Z2' dY4_Z2]; dY_M = [0 0 0 dY2_Z3' dY3_Z3' dY4_Z3]; dY_AR = [dY1_Z4' dY2_Z4' 0 0 0]; dY_Lambda = [dY1_Z5' dY2_Z5' 0 0 0 0]; dY_Sref = [dY1_Z6' dY2_Z6' 0 0 0 0];

REPORT DOCUMENTATION PAGE Form Appro OMB No. 0704 Public reporting burden for this collection of information is estimated to average 1 hour per response, including the time for reviewing instructions, searching exist										
gamening and maintaining the data needed collection of information, including suggesti Davis Highway, Suite 1204, Arlington, VA 2	, and completing and reviewing the collection of ons for reducing this burden, to Washington H 22202-4302, and to the Office of Management	r response, including the time for rev of information. Send comments rega eadquarters Services, Directorate for and Budget, Paperwork Reduction P	iewing instructions, searching existing data sources, ding this burden estimate or any other aspect of this Information Operations and Reports, 1215 Jefferson roject (0704-0188), Washington, DC 20503.							
1. AGENCY USE ONLY (Leave bla	ank) 2. REPORT DATE August 1998		PE AND DATES COVERED Memorandum							
4. TITLE AND SUBTITLE		rectificat	5. FUNDING NUMBERS							
Bi-Level Integrated Syst	tem Synthesis (BLISS)									
			509-10-31-03							
6. AUTHOR(S) Jaroslaw Sobieszczansk Sandusky, Jr.	i-Sobieski, Jeremy S. Agte, a	nd Robert R.								
7. PERFORMING ORGANIZATION	NAME(S) AND ADDRESS(ES)		8. PERFORMING ORGANIZATION REPORT NUMBER							
NASA Langley Research Hampton, VA 23681-21			L-17778							
9. SPONSORING/MONITORING A	GENCY NAME(S) AND ADDRESS(ES	\$)	10. SPONSORING/MONITORING AGENCY REPORT NUMBER							
National Aeronautics and Washington, DC 20546			NASA/TM-1998-208715							
11. SUPPLEMENTARY NOTES Sobieski: Manager, Computational AeroSciences, and Multidisciplinary Research Coordinator, NASA Langley Research Center; Agte: Graduate Research Scholar Assistant, The George Washington University, LT, U.S. Air Force; Sandusky: Professor, The George Washington University. 12a. DISTRIBUTION/AVAILABILITY STATEMENT Unclassified-Unlimited Subject Category 05 Distribution: Standard Availability: NASA CASI (301) 621-0390										
optimization, having a re optimizations that may e autonomous and may be sensitivity data, producin method improves that de variables are frozen and t the local variable subdor variables. Optimum sen using MATLAB and iSIC business jet design, and a agreement with the bench organization and map we	ptimization of engineering sy elatively small number of des ach have a large number of lo conducted concurrently. Sub as design improvement in er- sign in iterative cycles, each of the improvement is achieved by nains. In step two, further im- sitivity data link the second so GHT programming software a detailed design of an electron	ign variables, from the ocal design variables. T system and system opti- ach iteration. Starting f cycle comprised of two by separate, concurrent, provement is sought in tep to the first. The me and tested on a simplified ic device. Satisfactory of Modularity of the meth gy of concurrent process	od is intended to fit the human sing.							
			A03							
17. SECURITY CLASSIFICATION OF REPORT	18. SECURITY CLASSIFICATION OF THIS PAGE	19. SECURITY CLASSIFIC OF ABSTRACT	ATION 20. LIMITATION OF ABSTRACT							
Unclassified	Unclassified	Unclassified								

NSN 7540-01-280-5500

Standard Form 298 (Rev. 2-89) Prescribed by ANSI Std. Z-39-18 298-102