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<http://aero-comlab.stanford.edu/>

Aerospace Computing Laboratory

Stanford
Scientific Computing Research
with
Potential Payoffs
to the
Boeing Company

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Presentation to Dr. Robert Krieger, President, Boeing Phantom Works
St. Louis, May 17, 2004

Intelligent
Aerodynamics Inc.

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Overall Goal of Our Research

Promote a Lean Design Process

- Reduced Human and Computational Costs
- Potential for Superior Designs

University Role

Expand the Knowledge Base which will
Enable Improved Designs

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Three-Pronged Strategy

- More Cost Effective Computer Hardware
- More Efficient Algorithms
- Automatic Shape Design



Presenters

Overview:

Antony Jameson

Part1: Sci-Station

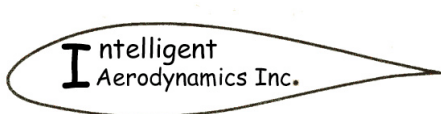
Gurjeet Singh

Part2: New Algorithms

Georg May

Part3: Shape Optimization

Kasidit Leoviriyakit



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Bios

Gurjeet Singh

Masters Student, Stanford University
B.Eng , Delhi University, India

2003-Present
2001

Georg May

Doctoral Candidate, Stanford University,
Dipl.Ing. Aachen Technical University, Germany
B.E Dartmouth College

2001-Present
2001
2000

Kasidit Leoviriyakit

Doctoral Candidate, Stanford University
M.S. Stanford University, Stanford University
B.Eng Kasetsart University, Thailand

2000-Present
1999-2000
1994-1998



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Section Shape Redesign of the Boeing 747

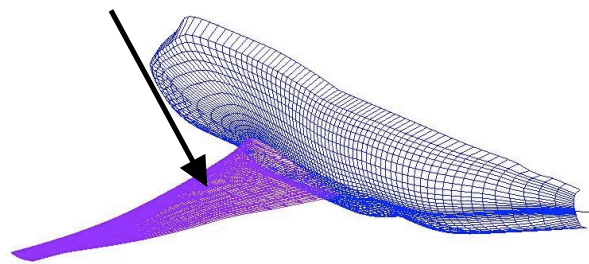
Baseline profile (red)

Redesigned profile (blue)



Profile at 50% span

Design Variables:
4224 points on the wing

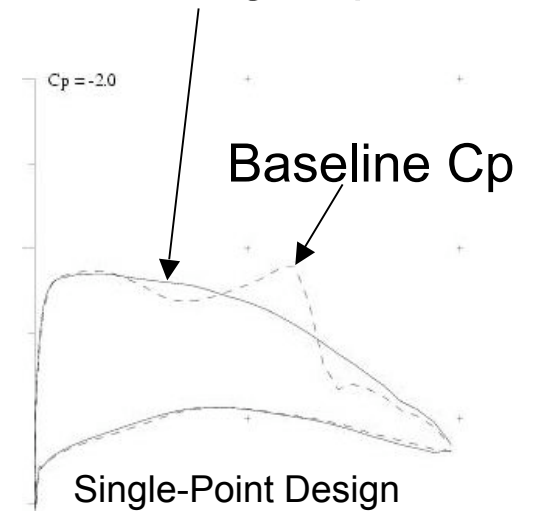


Boeing 747

Drag-Divergence Mach
Number Increased by 0.02

RANS simulation Re 100 million
3 design points - 30 design cycles
93 flow calculations
25000 sec with 8 - 1.7 GHz cpu
~ 340 trillion computer instructions

Redesign Cp

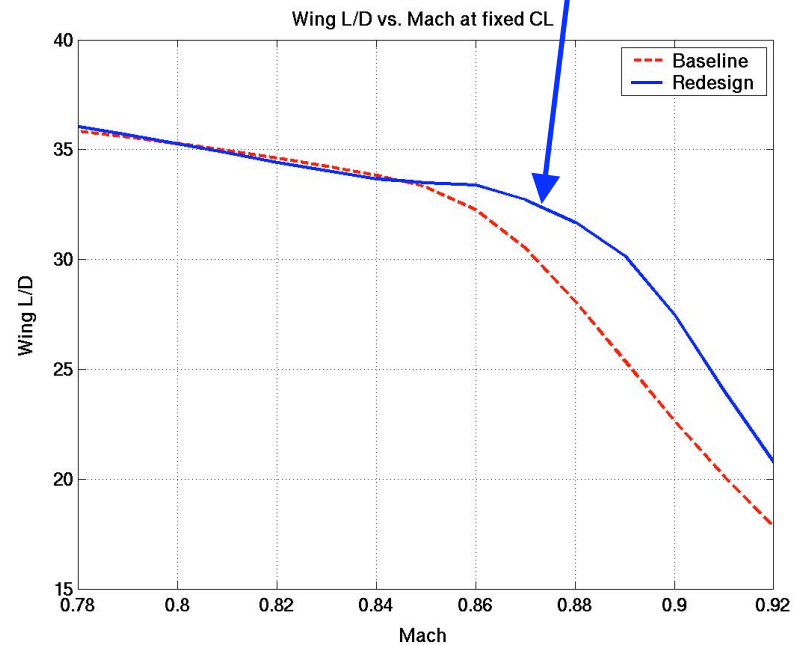
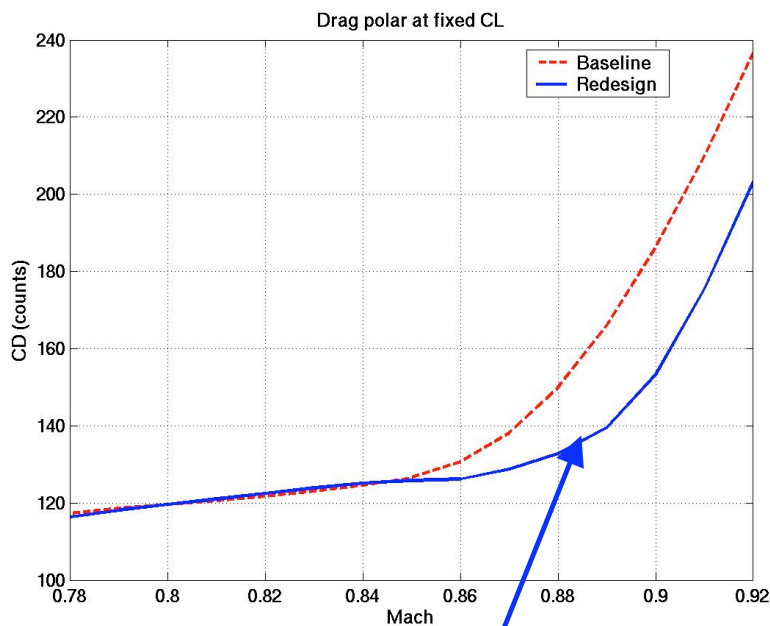


Redesign of the Boeing 747: Drag Rise

(Three-Point Design)

- Constraints
- : Fixed $C_L = 0.42$
 - : Fixed span-load distribution
 - : Fixed thickness

Improved wing L/D



Improved M_{DD}

benefit

benefit

Lower drag at the same Mach Number

Fly faster with the same drag



Executive Summary

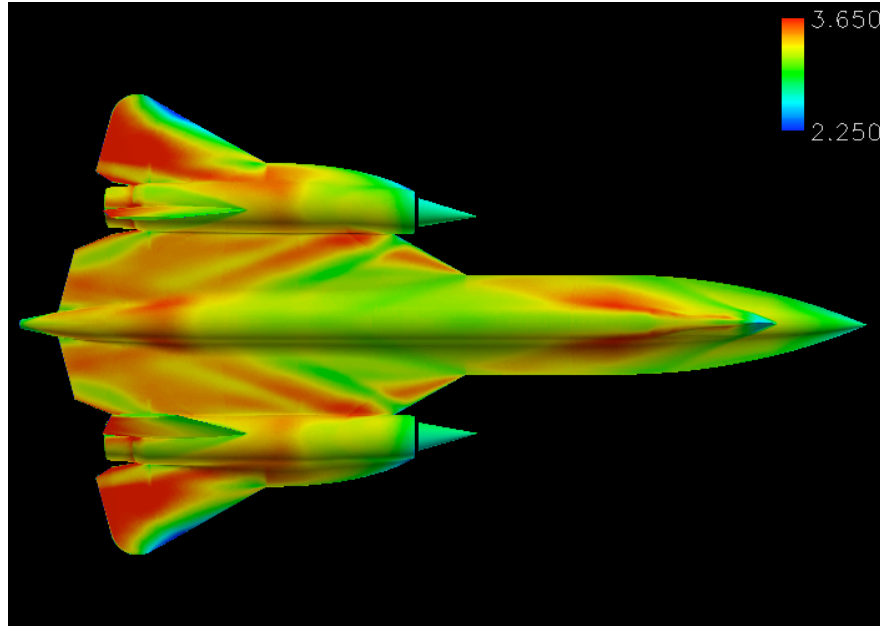
This presentation covers **three areas of research** which could strengthen the Boeing Company's engineering processes:

1. The design of a **"Sci-Station"** for scientific computation on the desktop to provide an order of magnitude increase in throughput
2. New **computational algorithms** which could provide an order of magnitude increase in throughput for steady flow simulations and two orders of magnitude for unsteady flow
3. Automatic **aerodynamic design** procedures based on control theory which could provide:
 - An order of magnitude reduction in human and computational costs
 - Potential for superior and unconventional designs
 - Freedom to design shapes as free surfaces with scalability to arbitrarily large numbers of design variables, and no need for user specified shape functions



Computation of SR71 with Flo3xx

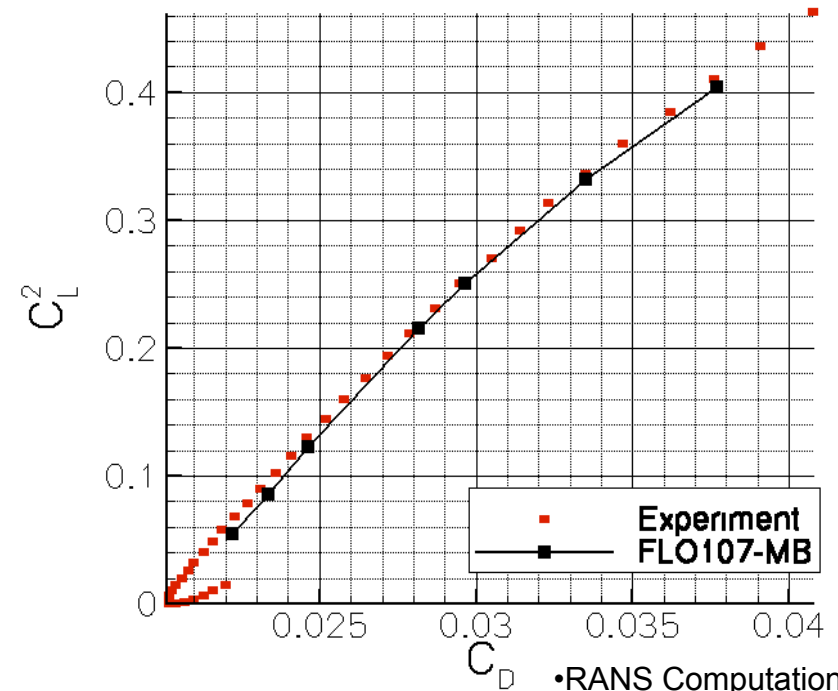
Mach Contours, CUSP scheme



- Euler Computation
- Geometry Courtesy of Lockheed Skunk Works

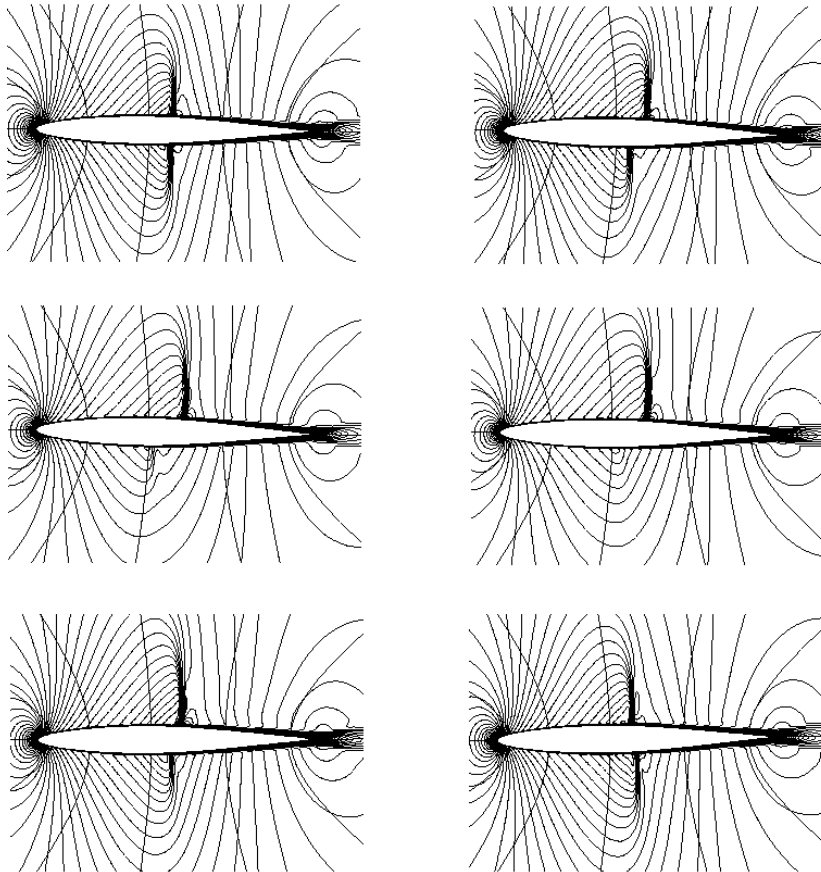
From IGES definition to completed result in one week, including CAD fixes, mesh generation

Validation of Flo107-MB for Drag Prediction Workshop on the DLR-F6 Configuration



- RANS Computation
- CUSP Scheme
- k-// Turbulence Model

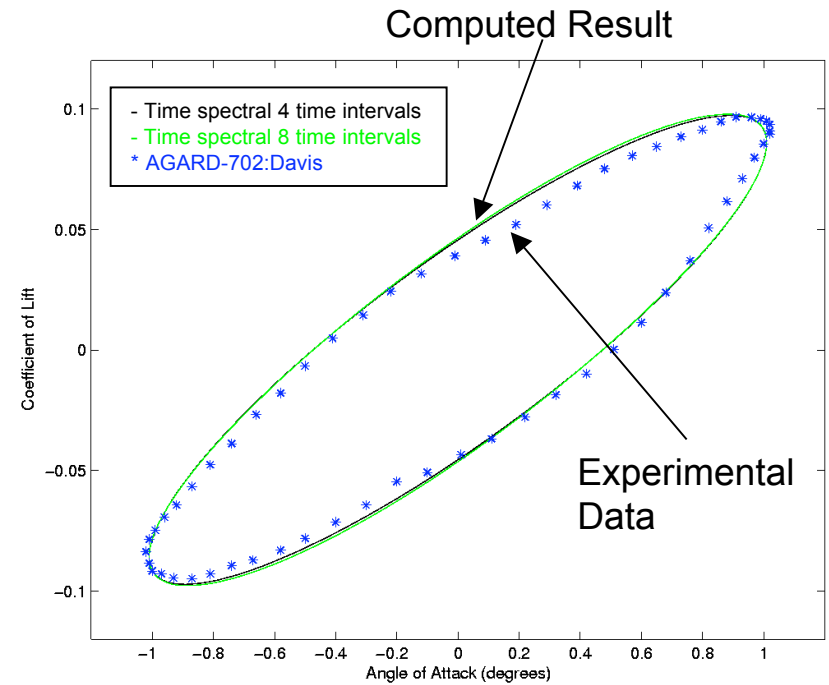




Pitching Airfoil (AGARD 702)
 Pressure Contours at Various
 Time Instances

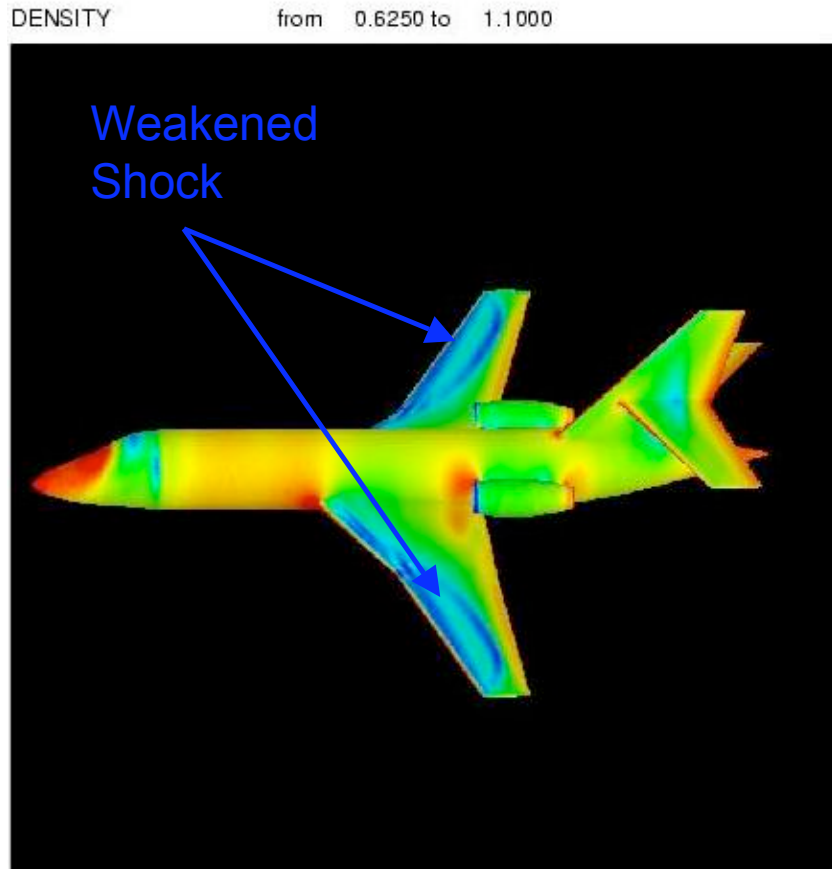
Results of Time Spectral Method

Periodic solution with 4 and 8 time intervals

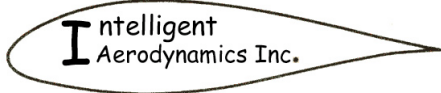


Automatic Redesign of Falcon Business Jet

Using SYNPLANE
Drag reduction 18 counts
at fixed $C_L = 0.4$



$C_D = 216$ counts



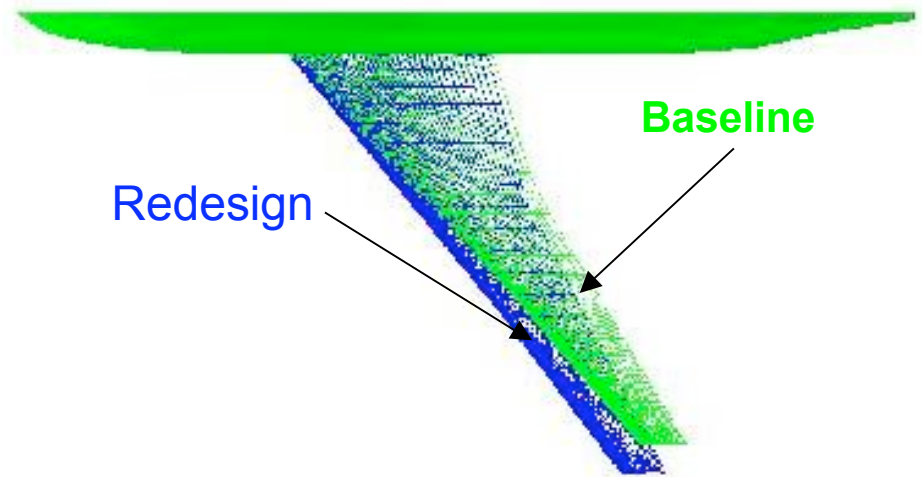
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Automatic Redesign of Boeing 747 Planform

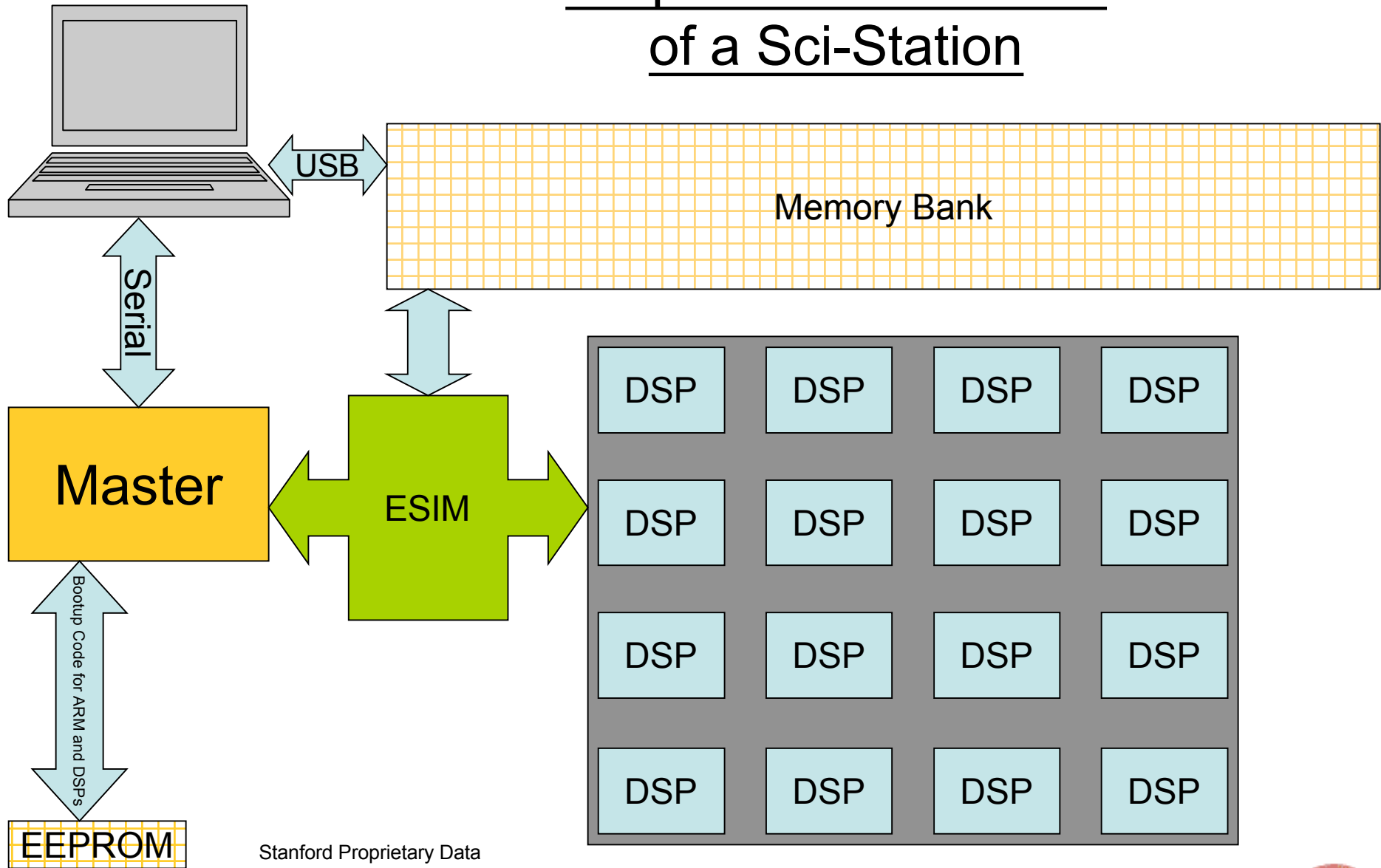
Using SYN88*
Drag reduction 21 counts
Weight reduction 800 pound
at fixed $C_L = 0.42$



$C_{D,wing} = 87$ counts
 $C_{w,wing} = 450$ counts



Proposed architecture of a Sci-Station



Stanford Proprietary Data

Propriety data

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Part 1

Concept of a SciStation

on each engineer's desktop

Presented by Gurjeet Singh

Goal : Enable all the calculations shown above to be routinely performed on the desktop

Payoff : Improved productivity in design and engineering

Patents pending

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Outline of the SciStation

- Need for specialized hardware
- Requirements
- Proposed architecture
- Steps of execution
- Programming
- Comparison with FPGA
- Comparison with ASIC
- Other applications
- Performance
- Limitations
- Resource requirements
- Proposal for work on a SciStation



Need for Specialized Hardware (1)

- Flow simulations are very time consuming
 - Current hardware is not optimized for scientific calculations
 - Most of crunching power is devoted to ‘peripheral’ computations
- Aerodynamic loads need eventually to be predicted for many thousands of points in the flight envelope
- We are moving towards aerodynamic shape optimization (ASO)
 - Each cycle of optimization may require multiple simulations
- Jet engine and helicopter wake simulations are still more complex



Need for Specialized Hardware (2)

- Increased computing power at the desktop level could enable engineers to obtain important data on the spot, and eliminate costs, management procedures and delays associated with a remote central computing system
- This could both accelerate the design process and increase productivity.



Need for Specialized Hardware (3)

- Current desktop equipment using general purpose processors (Athlon, Pentium etc.) have overtaken the previous generation of scientific workstations (Apollo, Sun, SGI etc), because of the dramatic increase in performance and decrease in cost of microprocessors, DRAM and disks.
- But there is a potential order of magnitude increase in desktop performance by optimizing machine's architecture for scientific computing (In which, the general microprocessor and PC manufacturers have little interest)
- A SciStation should be designed for a wide range of engineering applications:
 - Fluid mechanics
 - Solid mechanics
 - Heat transfer
 - Acoustics
 - Electromagnetics



Requirements

- For fluid mechanics, a system that can handle
 - 10 Million grid cells
 - 500 time steps
 - 4000 operations per time step
- Parallel decomposition of the problem
- Easy re-programmability
- Accuracy of simulation
- Speed



Proposed Architecture

- One master processor
- Several 'slave' processors
- One huge memory bank
- Interface with the PC
- Is the SciStation a co-processor to be added to a standard PC ?
 - YES (though its not a single chip co-processor)



Proposed Architecture

- Which slave processors to use ?
- Digital Signal Processors(DSP)
 - Originally designed for **real time applications**
 - Being used in **data and processing intensive tasks** such as digital cameras, MP3 players, nearly any smart electronic device
 - Extremely capable of handling **large amounts of data**
 - Extremely **fast** (because of real-time applications)
 - **Double precision accuracy without overhead**
 - Designed such because applications such as professional audio decoding and image processing etc. require high numerical precision
 - Technology scaling as Moore's law
 - Fastest processor available today : TI Cxx Raptor core
 - ~800Mhz, but if programmed optimally, can give max output ~3Ghz
 - IEEE Double precision compliant

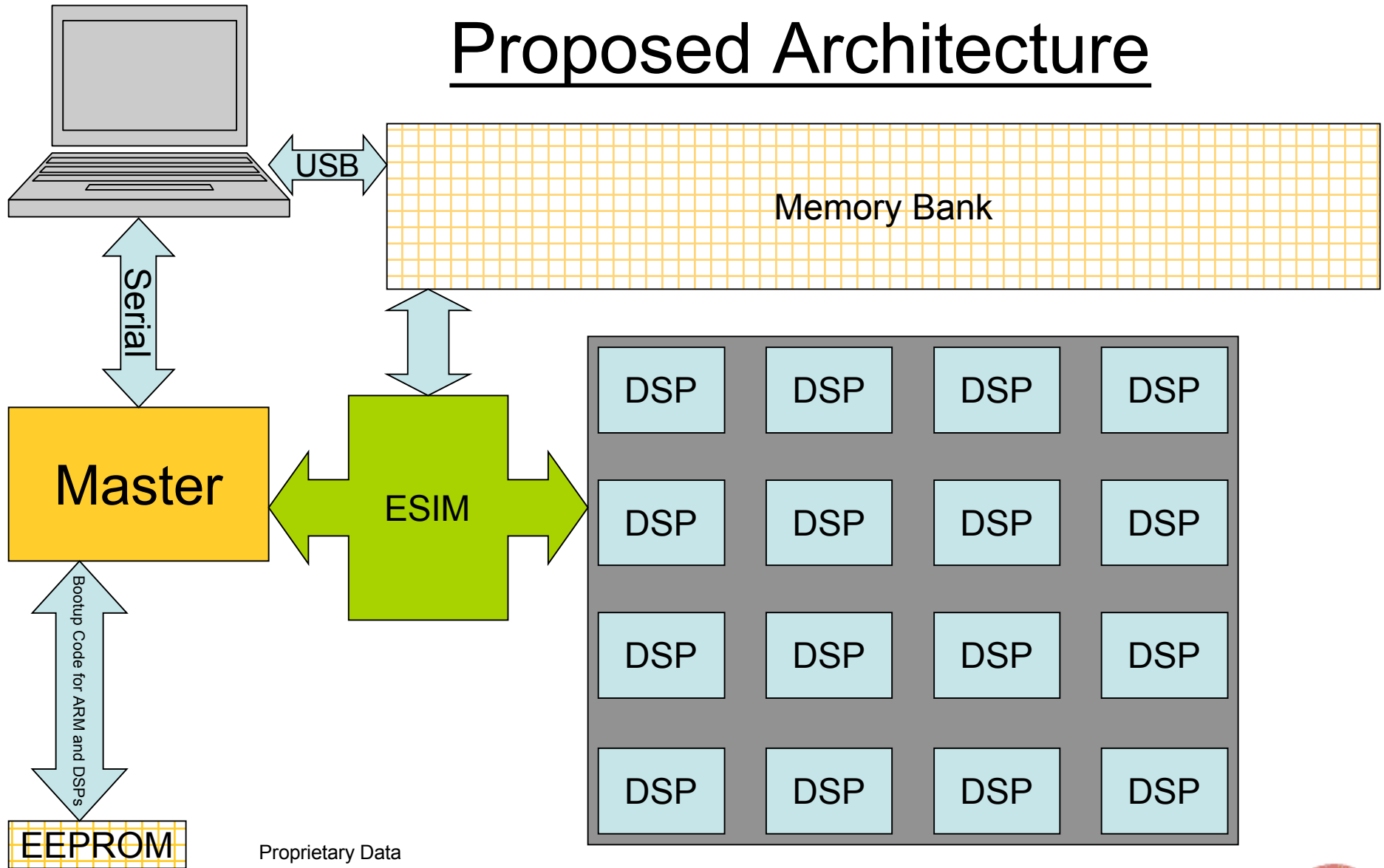


Proposed Architecture

- Master processor
 - Communicates with the PC
 - Fetches initial data
 - Pushes back the result
 - Delegates work packets to the DSPs.
 - Manages control signals and memory
- DSP
 - Computes solution ‘chunks’ as and when data is provided by the master processor.



Proposed Architecture



Patents pending

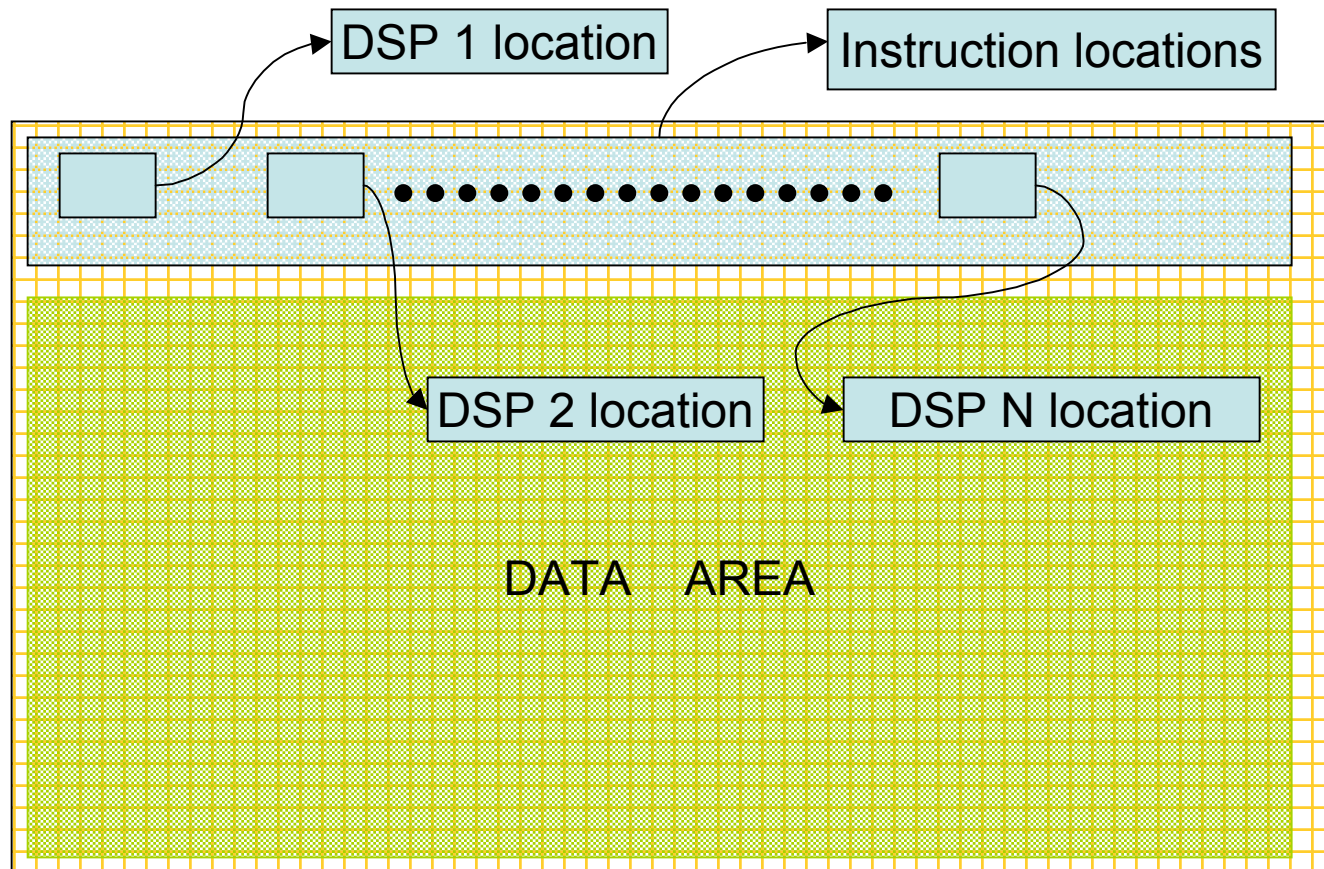


Proposed Architecture

- Shared memory parallelism
 - Each DSP has a fixed location and memory segment in the memory
 - The master processor installs data in the memory
 - The DSP's poll the memory for fresh data
 - The DSP's push results onto the memory



Proposed Architecture Shared Memory



Proprietary Data

Patents pending

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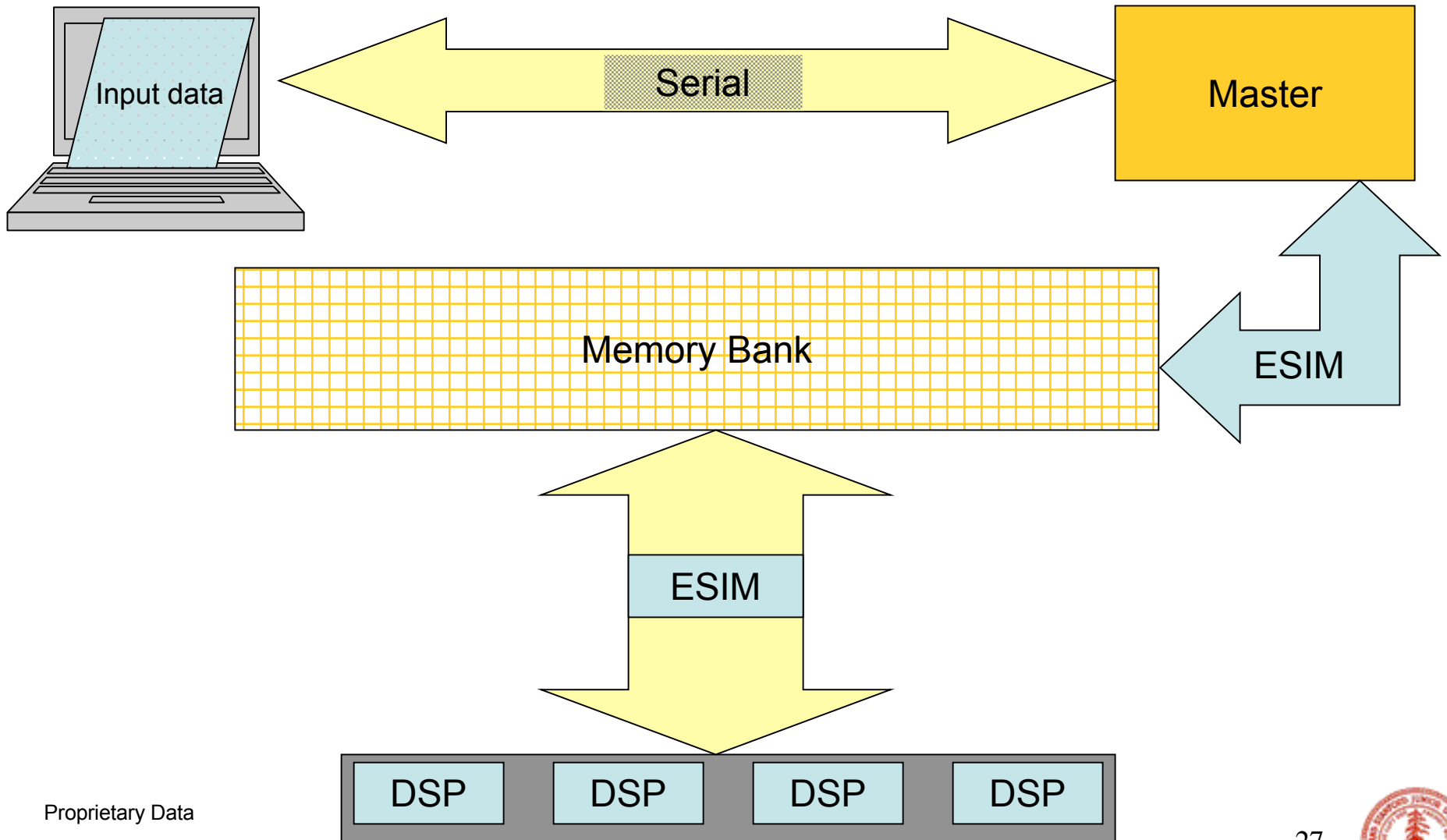


Steps of Execution

- Master processor receives data from the computer and stores it in the memory
- The master processor, burns code into each DSP from the memory
- DSP's poll the memory for their data
- DSP's complete computation and push the finished data back into the memory.



Steps of Execution



Proprietary Data

Patents pending

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Programming (Types of Codes)

- 2 – Tier programming
 - Code for each of the DSP's
 - Flow solver
 - Code installed at run-time by the master processor
 - Code for the Master processor
 - Parallelization schemes and data division
 - Communication and signal handling



Programming (Language Break-down)

- All the scientific code is in high level languages (such as C and Java), for easy re-programmability
- The basic operating code for the Master processor and DSP primitive polling code is in Assembly language. This code should not need to be changed too frequently



Programming (Code Break-down)

- Assembly language code
 - Used for bootstrap process
 - Find out system setup (such as number of processors)
 - Load up DSP code from memory
 - Load up Master processor code
 - Extremely light-weight and efficient
- High level language code
 - The main PDE solver
 - Loaded into DSP and Master memory by assembly code
 - Can be arbitrarily large



Comparison with Field Programmable Gate Arrays (FPGAs)

- Pros of FPGAs
 - Relatively easy to re-program
- Cons of FPGAs
 - Need to code ALU for any operation
 - Slower than DSPs
 - For parallelization, most of FPGA silicon will be engaged in IO instead of number crunching



Comparison with Application Specific Integrated Circuits (ASICs)

- Pros of ASICs
 - Extremely fast
 - Optimized performance
 - Make sense for large numbers
- Cons of ASICs
 - Not re-programmable
 - Made to order
 - Extremely expensive



Applications

- The architecture is good for parallelizable problems, in particular
 - Navier Stokes solver
 - Radar Cross-Section equations
 - FEM calculations
 - Any problem requiring solution to PDE's



Expected Performance using Currently Available Hardware

- Each DSP operates at ~800Mhz
 - Capability should track with Moore's law
- For
 - 1 Million grid cells
 - 500 time steps
 - 4000 operations per step
- One DSP takes 7 Hours
- Eight DSP's can do the work in an hour at max.



Limitations

- Memory access is not quick
 - Due to data access rates on inexpensive large memories(4-16GB)
 - Could be alleviated by local cache on each DSP
- Time required to shuttle data to and from PC may be a bottleneck
- The architecture is good for parallelizable problems, but less suited for serial problems
 - Could be alleviated by code kernels
- 32-bit data addressing
 - Could be alleviated by using a serial data access standard (e.g. USB 2/FireWire)



Resource Requirements

- Proof of concept with 3 DSP processors could be built by a single person (Knowledgeable of both computer hardware and numerical algorithms (Gurjeet Singh) (\$75,000 over 18 Months)
- 3 people might be needed to design a full prototype system with 16 processors
- Recurring Estimated Parts cost per unit : \$7700 for 16 processors



Proposal for Work on a SciStation

- **Phase 1**
 - Proof of Concept
 - Time : 18 Months
 - Funding : \$75,000
 - Gate Review
- **Phase 2**
 - Design of a prototype RANS SciStation
 - Time : 18 Months
 - Funding : \$225,000
 - Gate Review
- **Phase 3**
 - Prototype Assembly and port other problems
- **Phase 4**
 - Manufacturing





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Part 2

Enabling Algorithms in CFD

Embedded in a Unified Code Architecture:

Flo3xx

Presented by Georg May



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CFD Is Now on a Plateau

- Existing codes can reliably and accurately predict steady inviscid flow
- There are major issues in mesh generation, convergence, and accuracy for turbulent viscous flows over complex configurations, including prediction of transition and separation
- We aim to provide a better framework to tackle these issues



Overview

- A. **Flo3xx**: The latest addition to the widely used “FLO” series of codes
- B. **The Nonlinear SGS Multigrid Method**: Achieving textbook multigrid convergence
- C. **The BGK Method**: Using kinetic gas theory for aerodynamic problems of practical engineering interest
- D. **The Time-Spectral Method**: Time accurate solution of complex, periodically unsteady flow



Part 2A

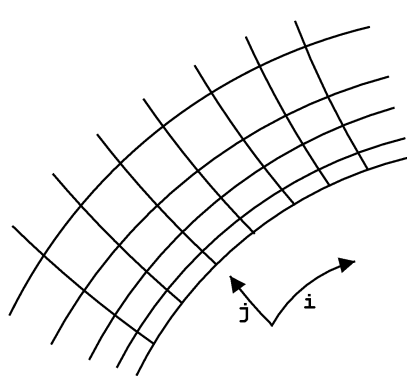
Flo3xx

Computational Aerodynamics on Arbitrary Meshes

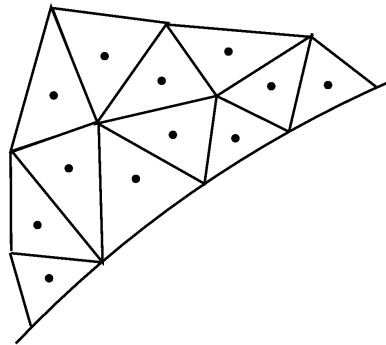


Support for Arbitrary Meshes

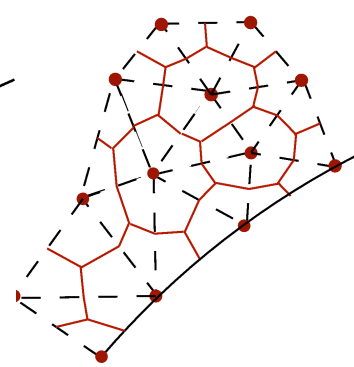
- Examples of mesh types which are being used in computational aerodynamics



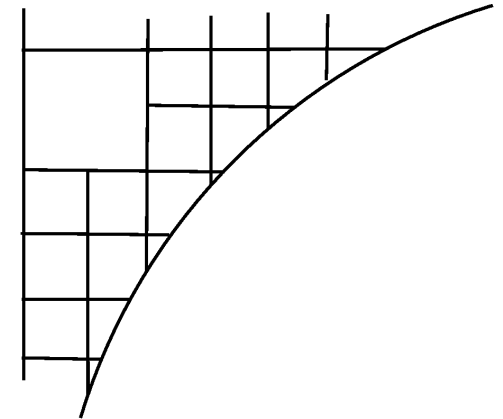
Structured



Unstructured
Cell-Centered



Unstructured
Cell-Vertex



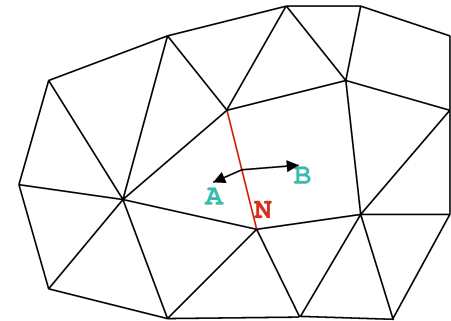
Nested Cartesian
With Cut Cells

- In **Flo3xx** a unified **mesh-blind** formulation supports all of these in one code
- Designed to meet the following objectives:
 - Platform for automatic mesh adaptation
 - Migration path to emerging mesh generation technologies
 - A robust algorithm that is tolerant to bad meshes



Support for Arbitrary Meshes

- Conservation laws are enforced on discrete control volumes
- Fluxes of conserved variables are exchanged through interfaces between these cells
- Independent of the mesh topology, each interface separates exactly two control volumes (on the right, face **N** separates cells **A** and **B**)



All algorithms are expressed in terms of a generic interface-based data structure



Treatment of Structured Meshes



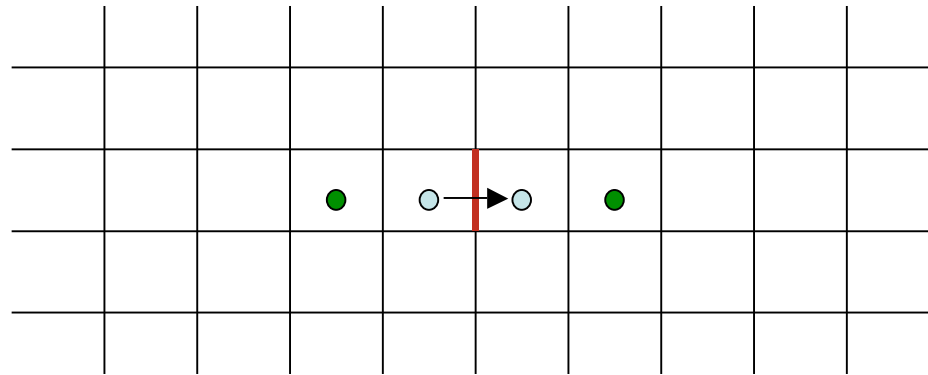
Interface Flux



First Neighbors



Second Neighbors

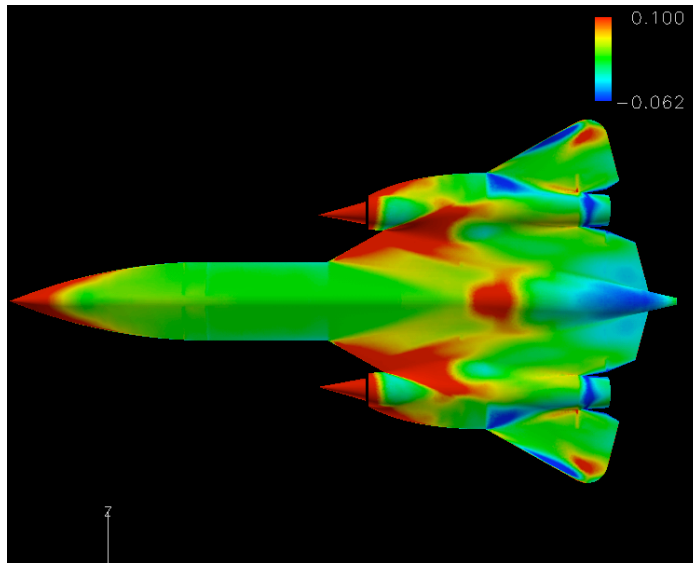


- Associate first and second neighbors with each face
- Allows implementation of standard schemes with five-point stencils (Jameson-Schmidt-Turkel JST, SLIP) in the same code
- Eliminates the need for gradient reconstruction
- Numerical experiments verify **25% overhead** due to indirect addressing in comparison with standard structured-code implementation (FLO107)

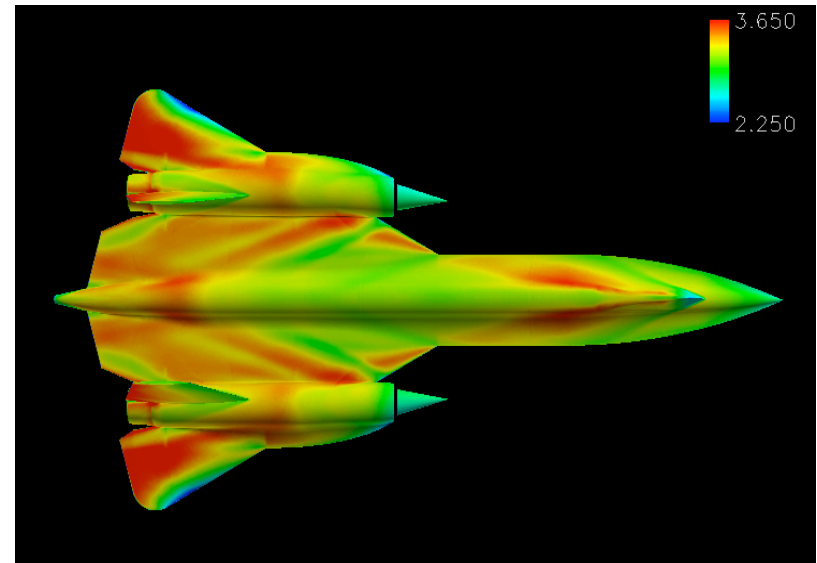


Flo3xx in Action...

Pressure Coefficient - Lower



Mach Number - Upper



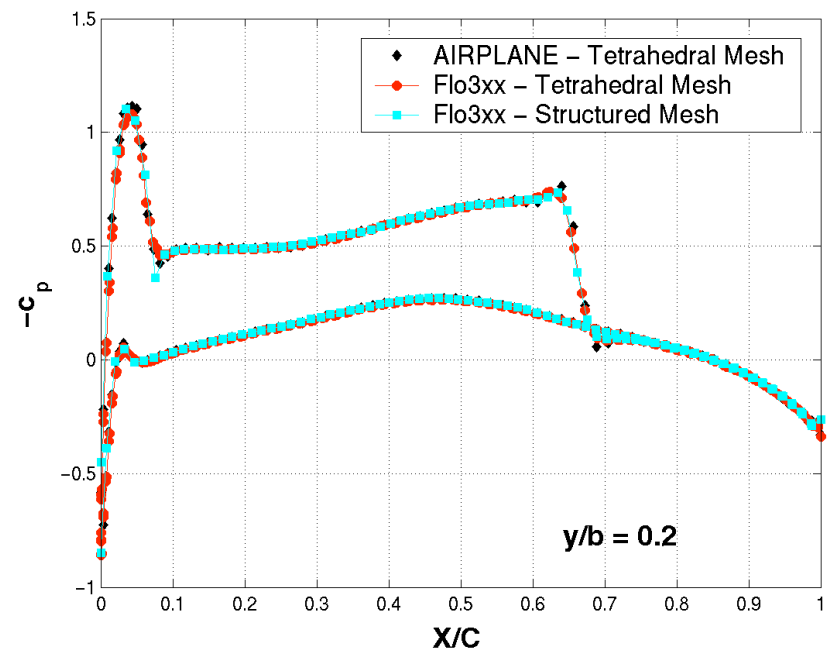
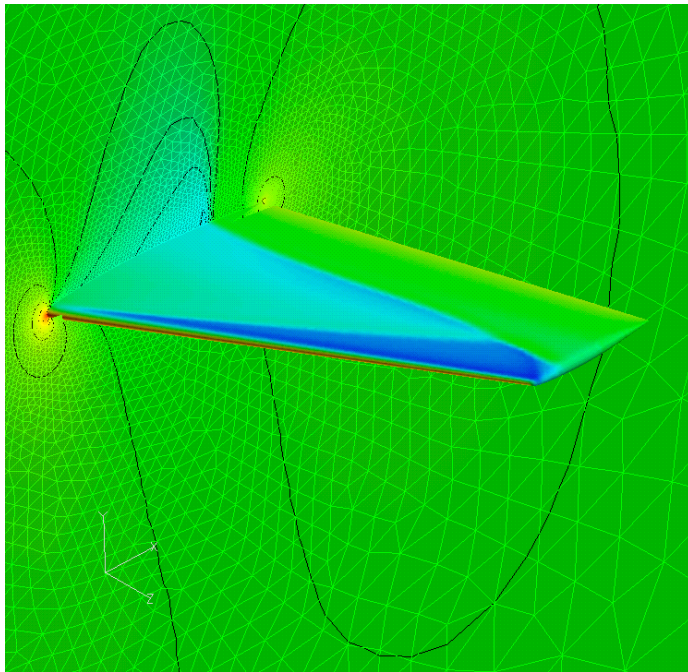
Geometry Courtesy of Lockheed Skunk Works

Lockheed SR71 at $M=3.2$, $\alpha=5$ deg. - Euler calculation with 1.5 Million grid points

- From IGES definition to completed result in one week, including CAD fixes, mesh generation
- We need to be able to compute extreme test cases
- This concerns both complexity of geometry and flow conditions

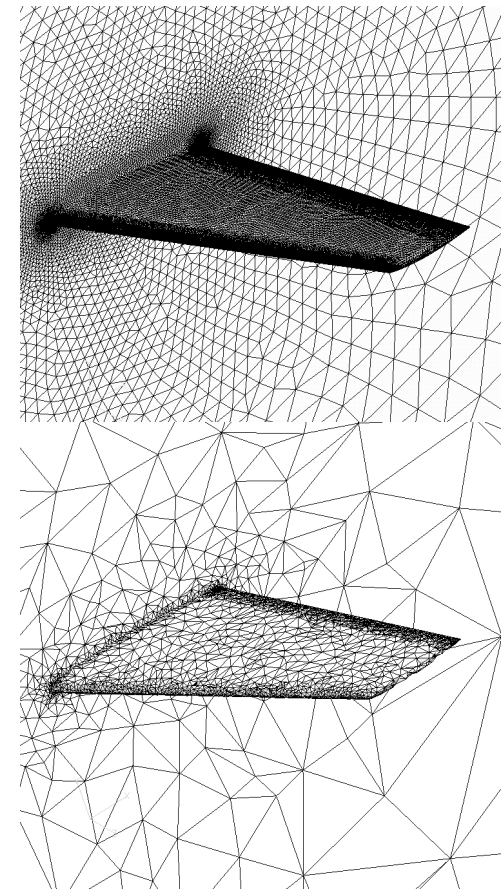
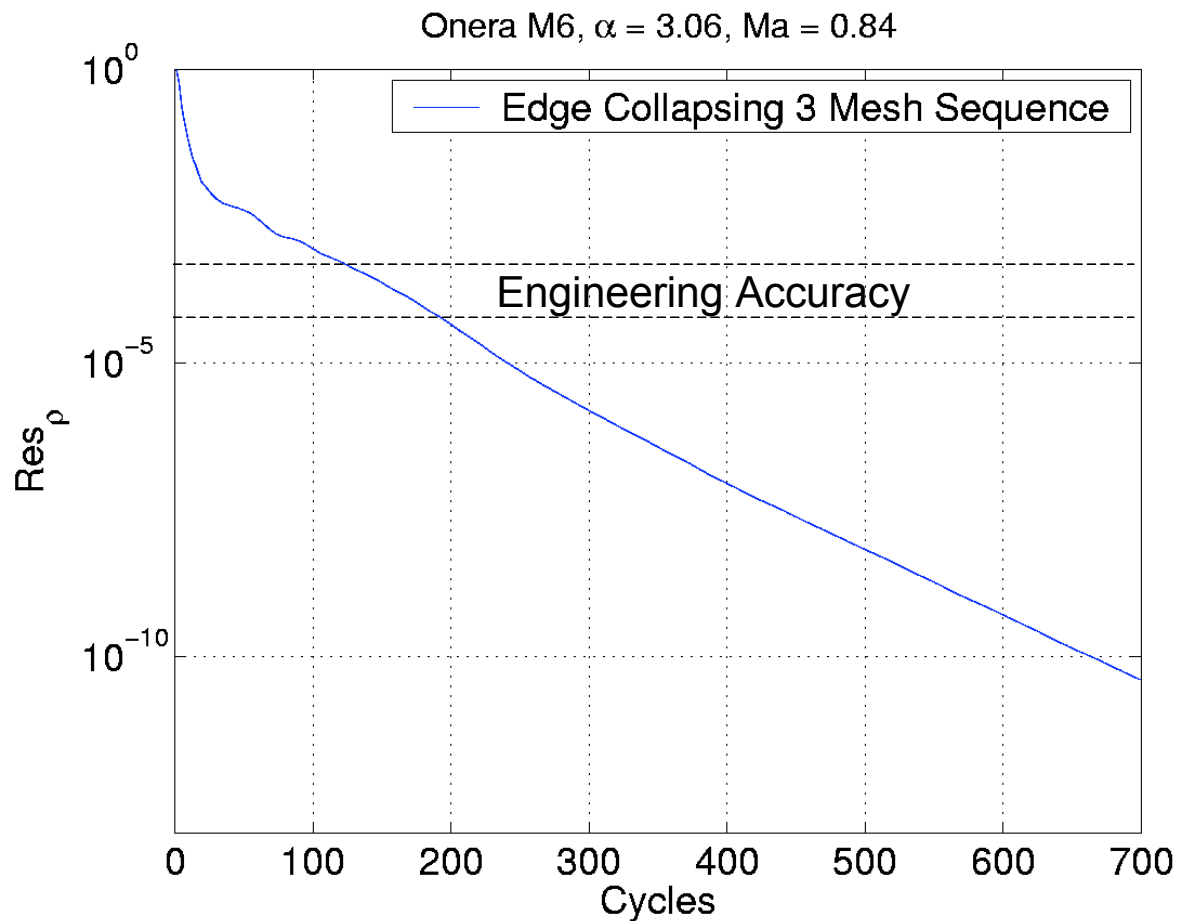


Validation of Unified Solution Algorithm in Flo3xx: Inviscid Transonic Flow

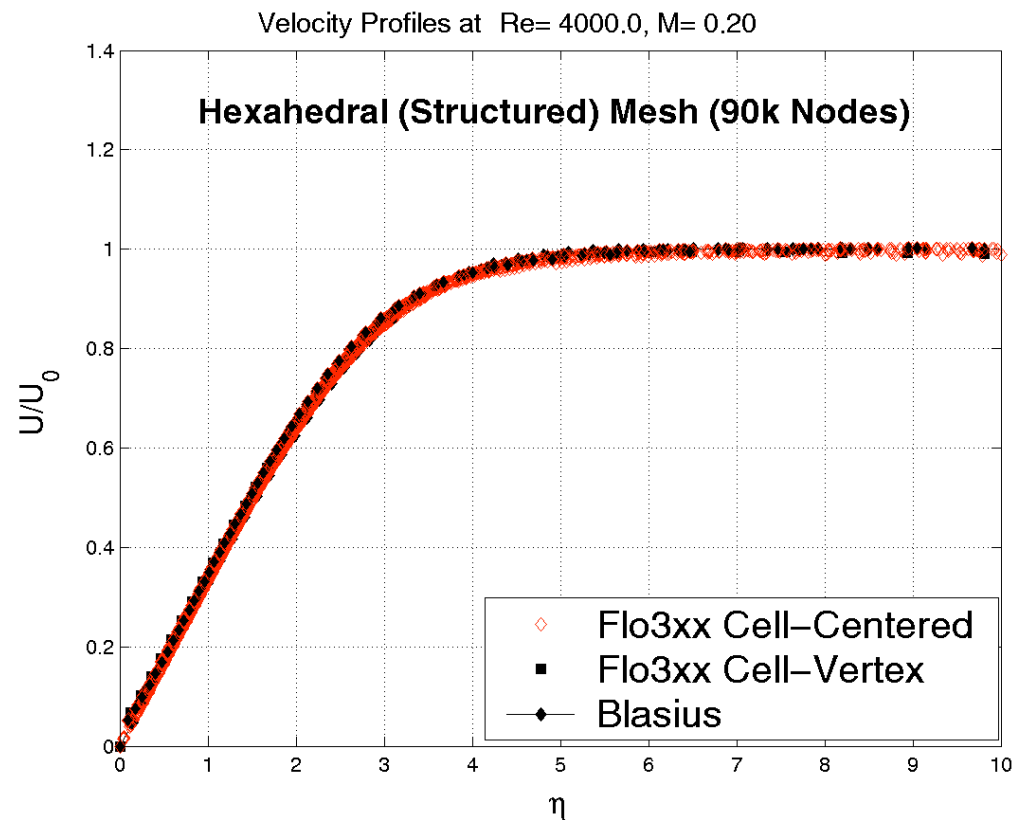


- Onera M6 Wing at $M=0.84$ and $\alpha = 3.06$ degrees

Convergence Using Automatic Multigrid

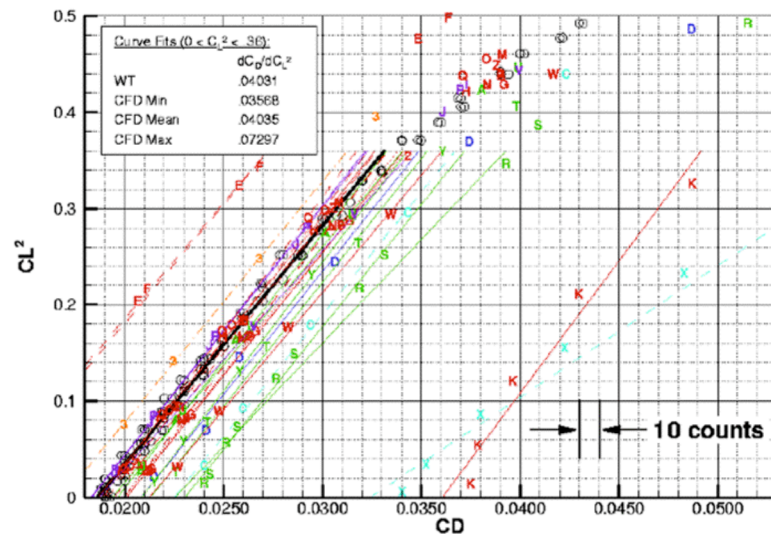


Initial Validation for Viscous Flow: Zero-Pressure-Gradient Boundary Layer

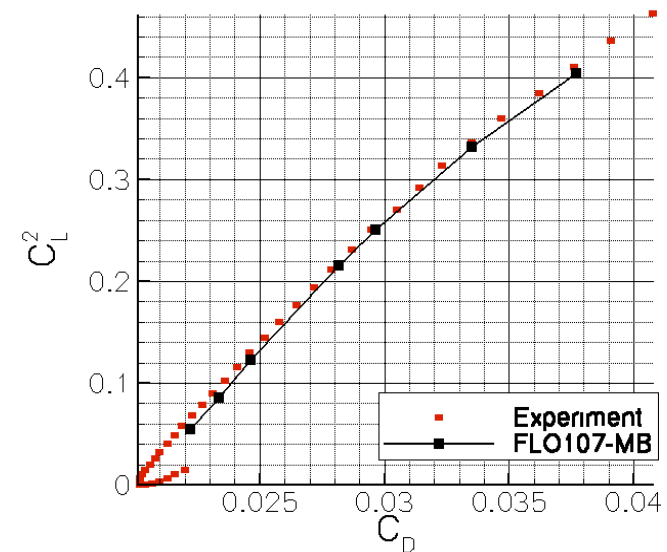


RANS Results Using FLO107-MB For Drag Prediction Workshop

Statistical Evaluation DPW1 – All Participants



Flo107-MB (DPW2)



- Accurate drag prediction for complex geometries in transonic flow is still very hard
- Flo3xx is currently in viscous validation phase.
- FLO107-MB has been thoroughly validated.
- Results of right figure were obtained with **CUSP** scheme and k-// turbulence model



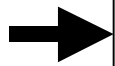
Flo3xx Payoffs

- **Highly flexible platform** for all applied aerodynamics problems and other problems governed by **conservation laws**
- **Fast turnaround** through convergence acceleration techniques
- Framework can be used to support advanced research, such as the **BGK method** or the **Time-Spectral Method**, which will be addressed in this talk
- This means, take advanced research out of a **laboratory setting** and apply it to problems of **practical engineering interest**, which is ultimately the only way to make an impact on the state-of-the art



Overview

A. **Flo3xx**: The latest addition to the widely used “FLO” series of codes



B. **The Nonlinear SGS Multigrid Method**: Achieving textbook multigrid convergence

C. **The BGK Method**: Using kinetic gas theory for aerodynamic problems of practical engineering interest

D. **The Time-Spectral Method**: Time accurate solution of complex, periodically unsteady flow



Part 2B

Non Linear Symmetric Gauss-Seidel Multigrid Scheme

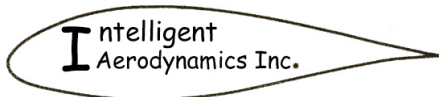
Jameson + Caughey 2001

Evolved from LUSGS scheme

Yoon + Jameson (1986)

Rieger + Jameson (1986)

Achieved “Text Book” Multigrid Convergence



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Nonlinear Symmetric Gauss-Seidel (SGS) Scheme

Forward and reverse sweeps:

For 1D case:
$$\frac{\partial w}{\partial t} + \frac{\partial}{\partial x} f(w) = 0$$

Sweep (1): Increasing j \longrightarrow

$$w_j^{(1)} = w_j^{(0)} - \frac{1}{|A|} \left(f_{j+\frac{1}{2}}^{(00)} - f_{j-\frac{1}{2}}^{(10)} \right)$$

$$f_{j-\frac{1}{2}}^{(01)} = f\left(w_j^{(0)}, w_{j+1}^{(1)}\right)$$

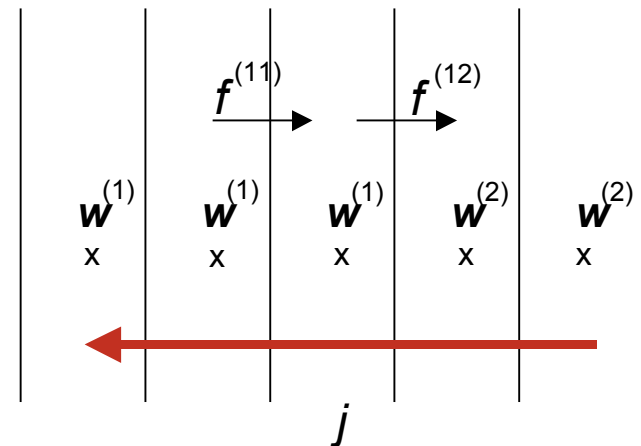
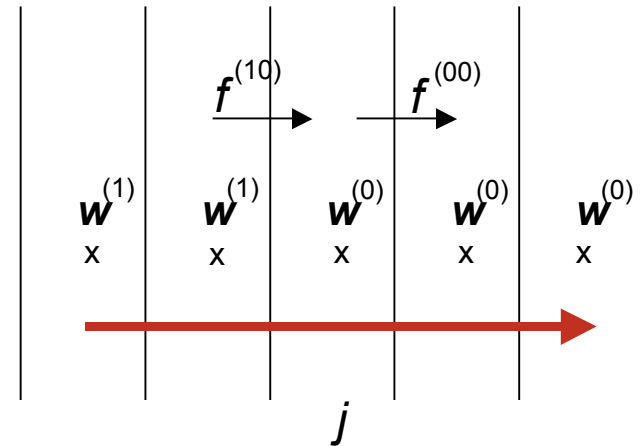
$$A = \frac{\partial f}{\partial w}$$

Sweep (2): Decreasing j \longrightarrow

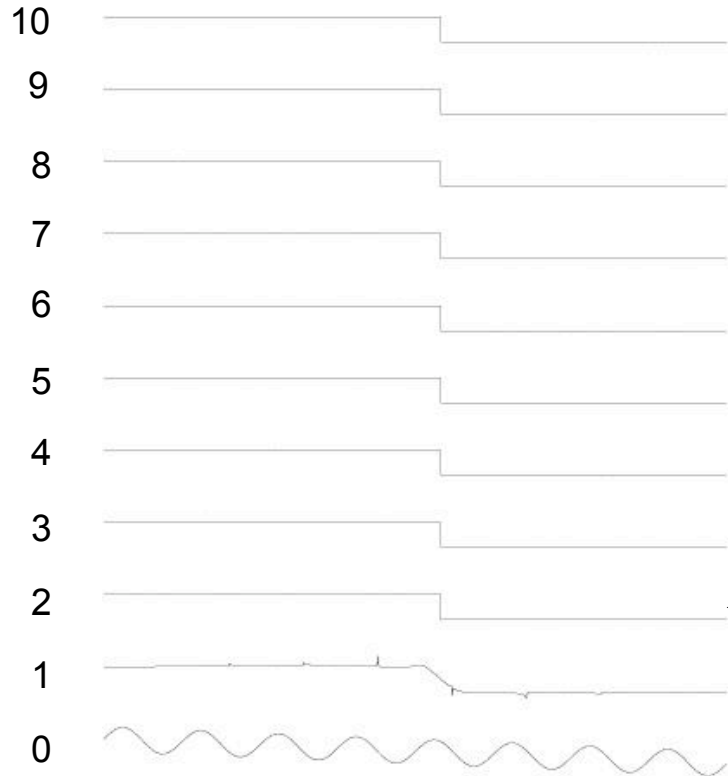
$$w_j^{(2)} = w_j^{(1)} - \frac{1}{|A|} \left(f_{j+\frac{1}{2}}^{(12)} - f_{j-\frac{1}{2}}^{(11)} \right)$$

4 Flux evaluations in each double sweep

Cost per iteration similar to 4 - stage Runge - Kutta scheme



Solution of Burgers Equation on 131,072 Cells in Two Steps With 15 Levels of Multigrid



SOLUTION OF UT +UUX = 0.

NX N MESH
131072 15

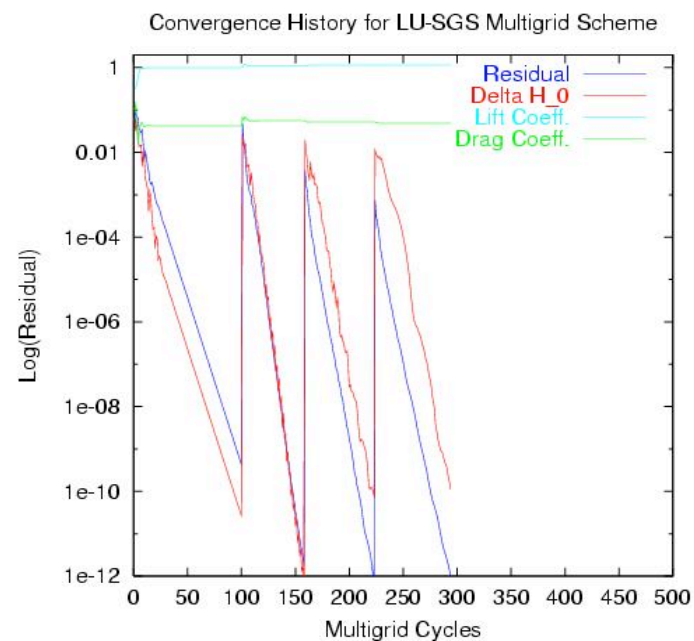
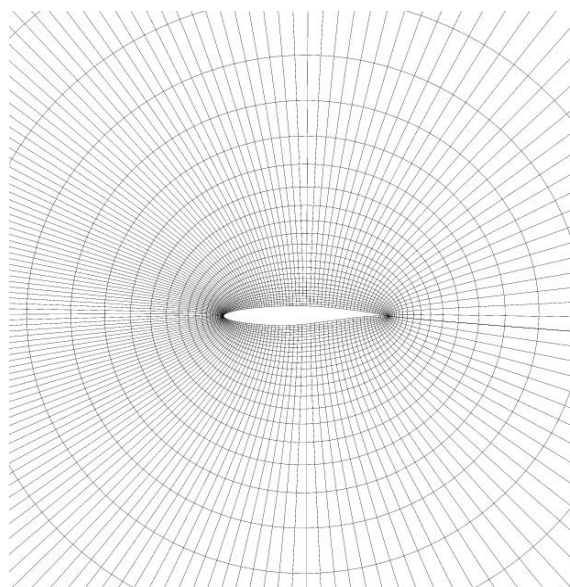
CFL
1.0000

CYCLE	AVG RESIDL	MAX RESIDL	IMAX
1	0.8038E-02	0.1295E+01	57667
2	0.3413E-04	0.4337E+00	70962
3	0.2773E-06	0.9391E-01	70961
4	0.1576E-14	0.9567E-05	70959
5	0.3037E-28	0.7905E-13	70960
6	0.3037E-28	0.7905E-13	70960
7	0.3037E-28	0.7905E-13	70960
8	0.3037E-28	0.7905E-13	70960
9	0.3037E-28	0.7905E-13	70960
10	0.3037E-28	0.7905E-13	70960

SOLUTION OF BURGERS EQUATION BY SYMMETRIC RELAXATION
131072 CELLS 15 LEVELS
CFL 1.000 RAVG 0.0



Solution of 2D Euler Equations: Convergence for NACA0012

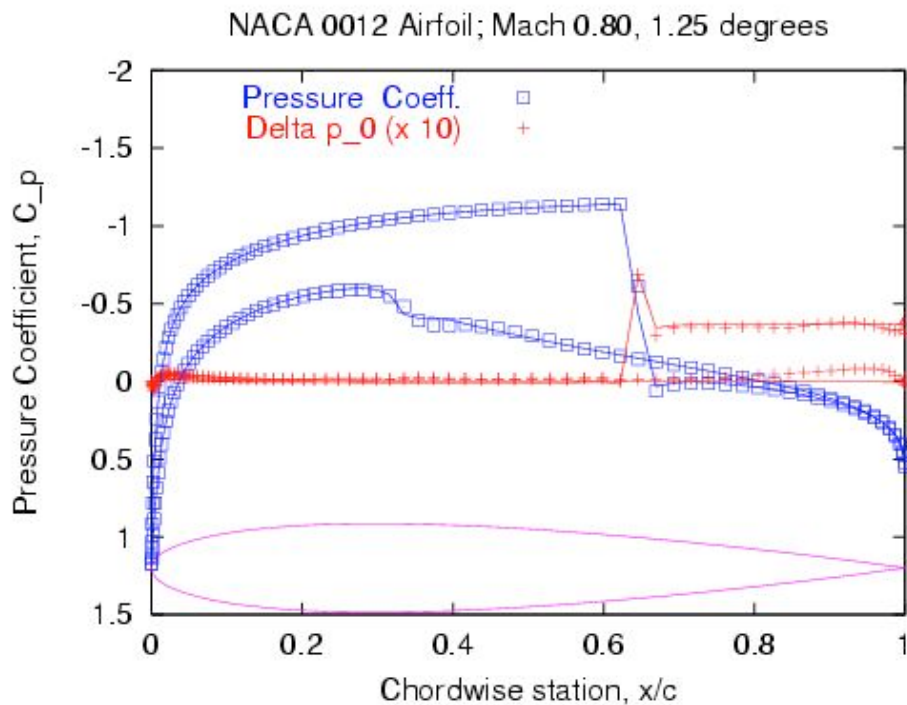


- The convergence history shows the successive computation on meshes of different sizes
- The convergence rate is independent of the mesh size
- Convergence rate $\sim .75$ per cycle

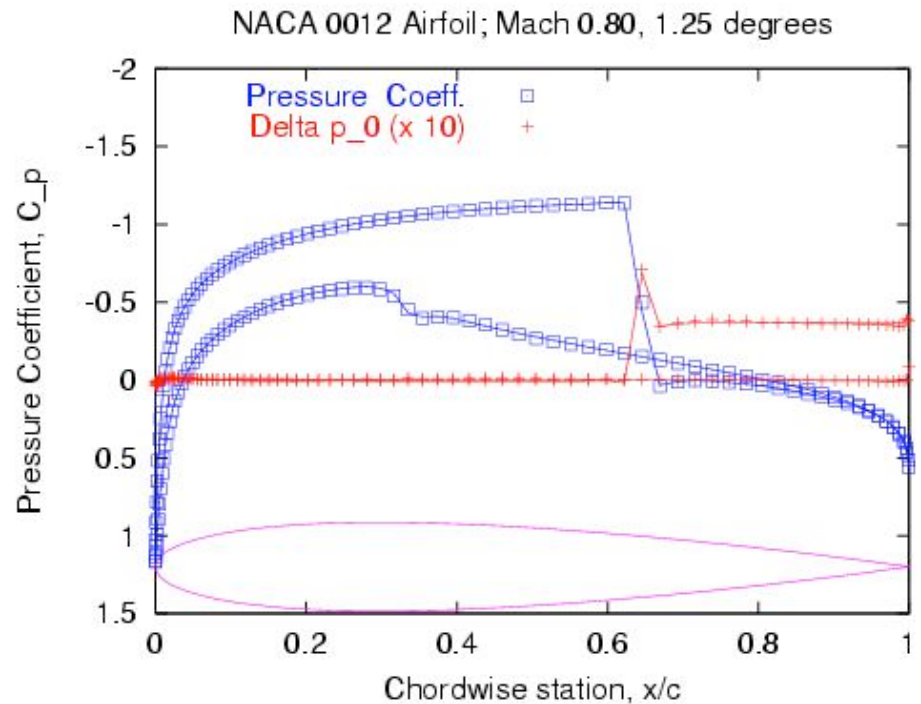


Solution of 2D Euler Equations

NACA0012 Airfoil



Solution after 3 multigrid cycles



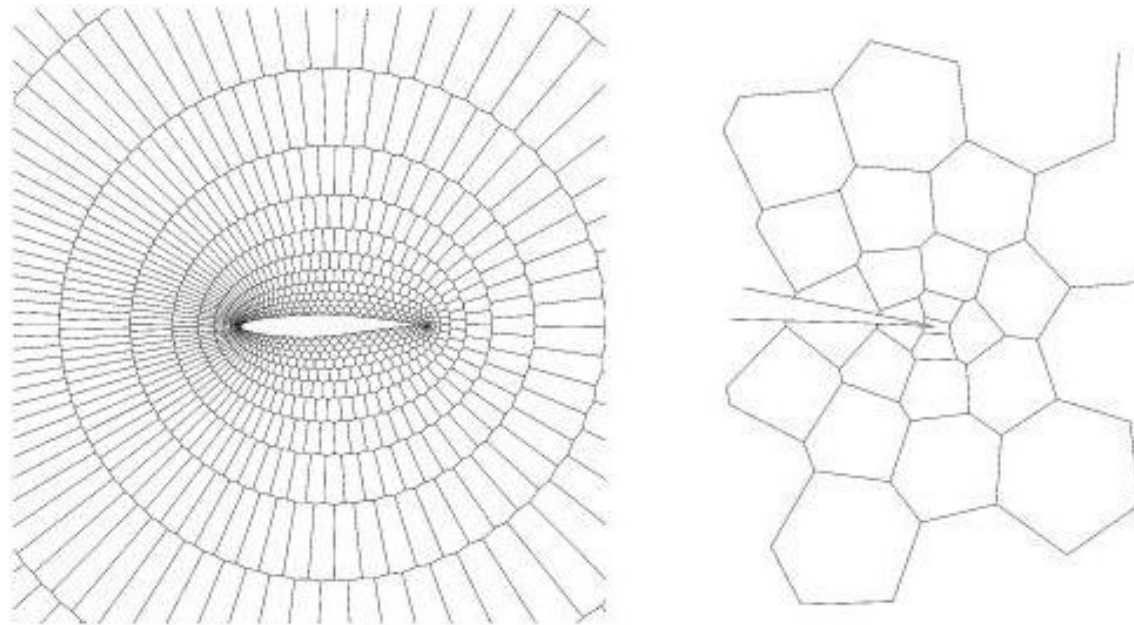
Solution after 5 multigrid cycles

Solid lines: fully converged result



Face-based Gauss Seidel (FBGS) Scheme

(Following a suggestion by John Vassberg)



- On an arbitrary grid, loop over faces instead of looping over cells
- Update the cells adjacent to a face as you go along
- Updated state will be used on next visit to a cell



Payoff of Fast Convergence (Solution in 10 - 300 Iterations)

- Faster turnaround
- Increased throughput
- Increased productivity
- Improved accuracy
- Increased reliability

Note: Gulfstream engineers report that the **WIND code** needs **30,000 iterations** to produce a converged RANS solution for a supersonic inlet



Overview

A. **Flo3xx**: The latest addition to the widely used “FLO” series of codes

B. **The Nonlinear SGS Multigrid Method**: Achieving textbook multigrid convergence

→ C. **The BGK Method**: Using kinetic gas theory for aerodynamic problems of practical engineering interest

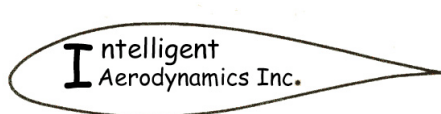
D. **The Time-Spectral Method**: Time accurate solution of complex, periodically unsteady flow



Part 2C

The Finite-Volume BGK Scheme

Using Statistical Mechanics to Enhance
Computational Aerodynamics



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60



A Major Conceptual Difference Between Continuum Mechanics and Statistical Mechanics

- In continuum mechanics the unknown solution variables are defined “pointwise” with precise values:

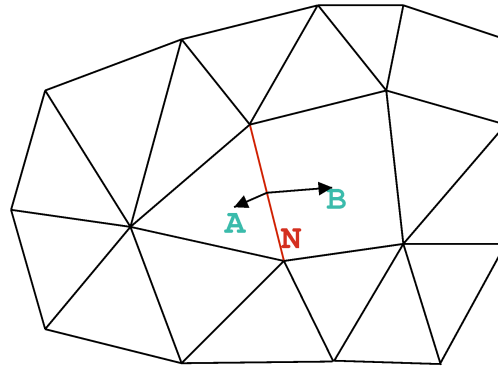
$$U = U(x, y, z, t)$$

- In statistical mechanics the solution variables exist only as moments of a statistical distribution in physical and phase space, or as “expectation values”:

$$U = \int u f(x, y, z, u, v, w, \dots, t) du dv dw d\dots$$



The Key Idea of the Finite-Volume BGK Scheme



- Compute the fluxes for the Navier-Stokes equations at interface **N** from the distribution functions in cells **A** and **B**
- A time-dependent distribution function needs to be constructed at each time step for each cell



Finding the Distribution Function

- The equilibrium distribution function is known from Boltzmann statistics:

$$f_{eq} = g(x, y, z, u, v, w, \rho) = A(x, y, z) e^{-\rho/(x, y, z) \{U^2 u^2 + (V^2 v^2 + (W^2 w^2 + \rho^2)\}}$$

- The nonequilibrium distribution function is unknown, but its evolution is given by the Boltzmann equation:

$$\frac{\partial f}{\partial t} + u \frac{\partial f}{\partial x} + v \frac{\partial f}{\partial y} + w \frac{\partial f}{\partial z} = Q(f, f)$$

Collision Integral

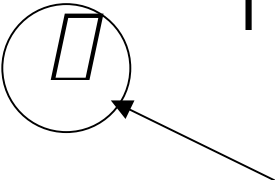
- Global numerical solution infeasible, because of high dimensionality



A Crucial Simplification (Bhatnagar, Gross & Krook - BGK)

- Replace the Collision Integral Q with a linear relaxation term:

$$Q = -\frac{f - g}{\tau} \quad || \quad \frac{\partial f}{\partial t} + u \frac{\partial f}{\partial x} + v \frac{\partial f}{\partial y} + w \frac{\partial f}{\partial z} = -\frac{f - g}{\tau}$$


 Collision Time

- This equation can be solved analytically:

$$f(\vec{x}, \vec{u}, t, \tau) = \int_0^t g(\vec{x} - \vec{u}(t - t'), \vec{u}, t', \tau) e^{-\frac{(t-t')}{\tau}} dt' + e^{-\frac{t}{\tau}} f_0(\vec{x} - \vec{u}t, \vec{u}, \tau)$$



A Key Observation

- By Chapman-Enskog expansion the Navier Stokes equations can be recovered from the BGK equation, with the viscosity coefficient

$$\mu = \mu p$$

- By setting the collision time τ appropriately, Navier-Stokes fluxes can be computed directly from the distribution function



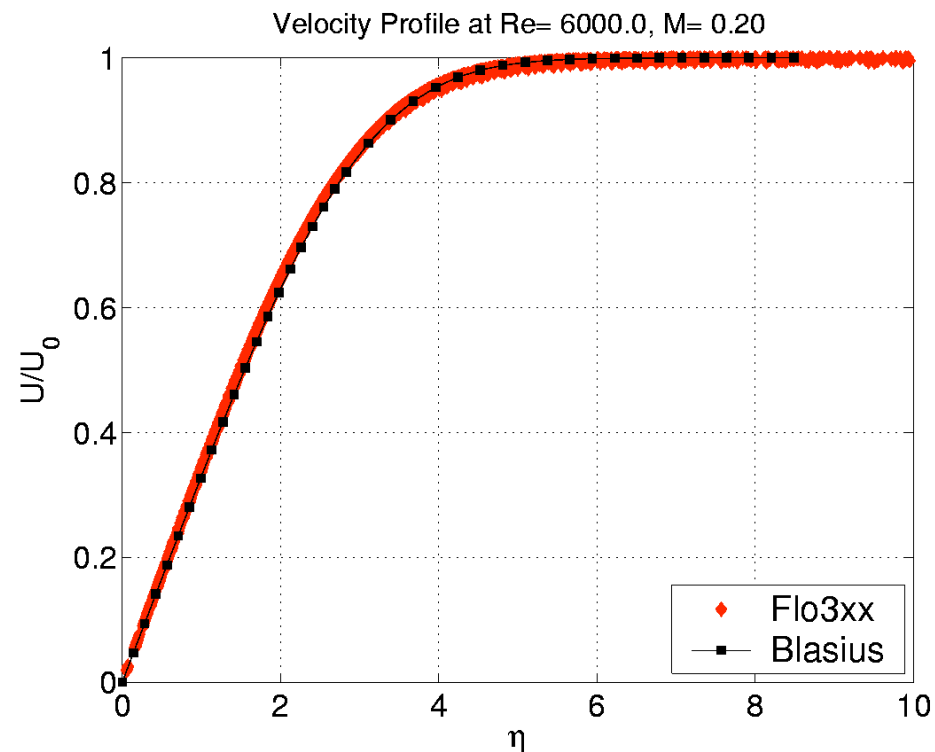
Payoff

- It is not necessary to compute the rate of **strain tensor** in order to calculate viscous fluxes
- This **eliminates** the need to perform **two levels of numerical differentiation**, which is difficult on arbitrary meshes
- Improved accuracy and reduced sensitivity to the quality of the mesh
- **Automatic upwinding via the kinetic model**, with no need for explicit artificial diffusion, thus reduced computational complexity



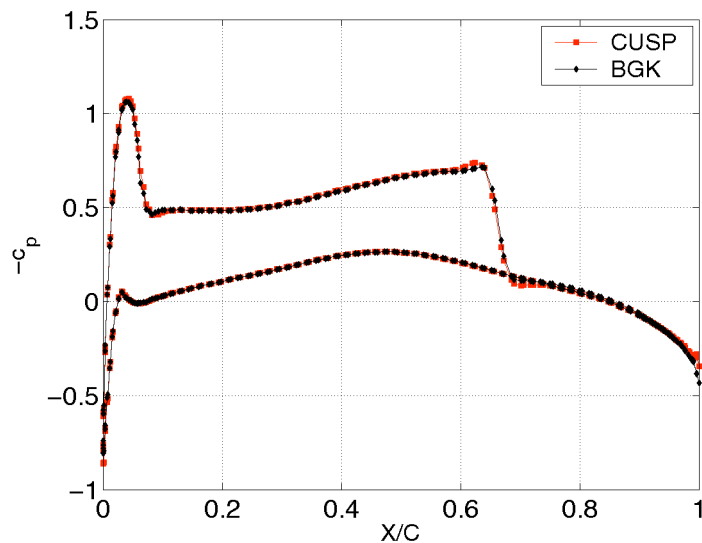
Validation of the BGK Scheme: Zero-Pressure-Gradient Boundary Layer

Blasius Profile

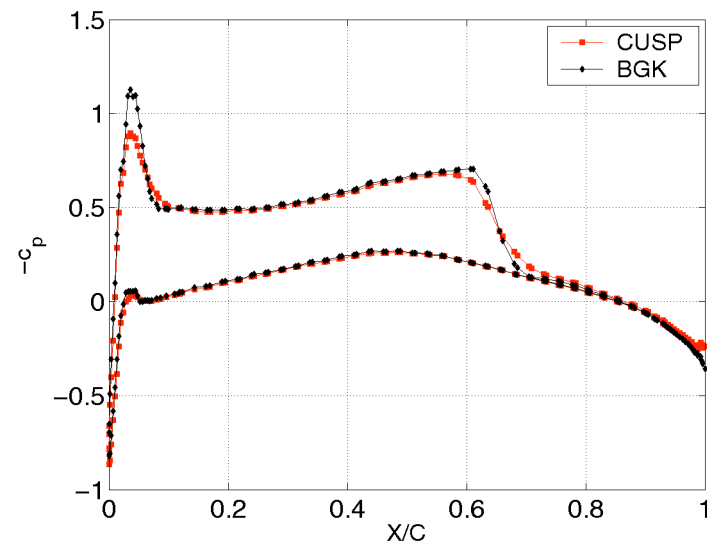


Validation of the BGK Scheme: Inviscid Transonic Flow

Finer Mesh (316k Nodes)



Coarser Mesh (94k Nodes)

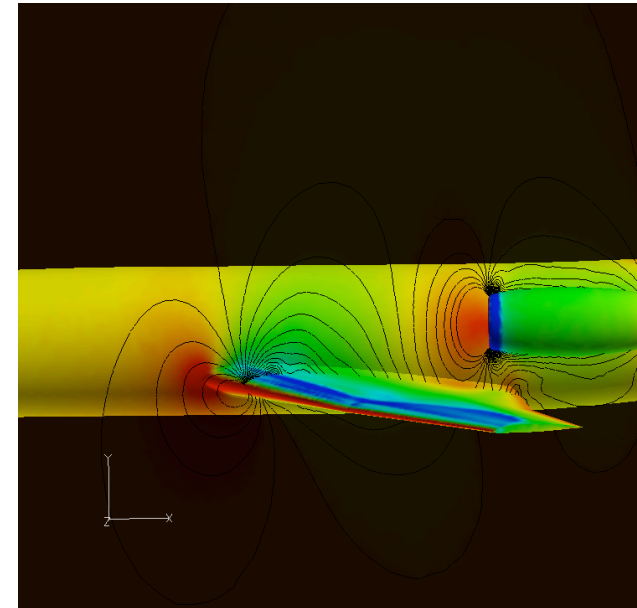
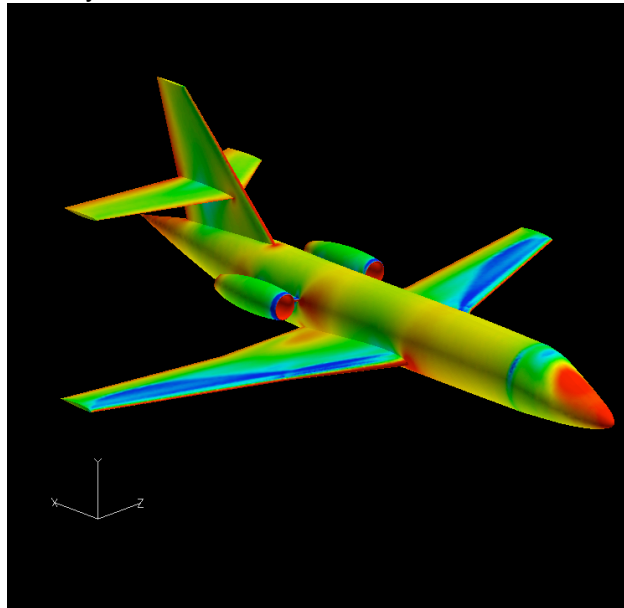


- Onera M6 Wing at $M=0.84$, $\alpha = 3.06$ degrees
- With sufficient resolution CUSP and BGK give similar results
- BGK seems to handle lower-resolution meshes better
- This might allow a reduction in the number of mesh points



Validation of the BGK Scheme using Flo3xx: Inviscid Transonic Flow

Density from 0.625 to 1.1



- Falcon Business Jet
- $M = 0.8$
- Angle of Attack: 2 degrees



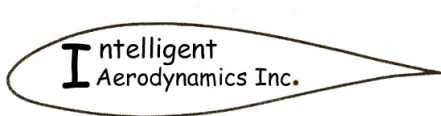
Overview

- A. **Flo3xx**: The latest addition to the widely used “FLO” series of codes
- B. **The Nonlinear SGS Multigrid Method**: Achieving textbook multigrid convergence
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- D. **The Time-Spectral Method**: Time accurate solution of complex, periodically unsteady flow



Part 2D

Fast Time Integration Methods for Unsteady Problems



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Potential Applications

- Flutter Analysis,
- Flow past Helicopter blades,
- Rotor-Stator Combinations in Turbomachinery,
- Zero-Mass Synthetic Jets for Flow Control



Dual Time Stepping BDF

The kth-order accurate **backward difference formula** (BDF) is of the form

$$D_t = \frac{1}{\Delta t} \sum_{q=1}^k \frac{1}{q} (\Delta t)^q \quad \text{where} \quad \Delta t w^{n+1} = w^{n+1} \Delta t w^n$$

The non-linear BDF is solved by inner iterations which advance in pseudo-time t^*

The second-order BDF solves

$$\frac{dw}{dt^*} + \frac{3w - 4w^n + w^{n+1}}{2\Delta t} + R(w) = 0$$

Implementation via

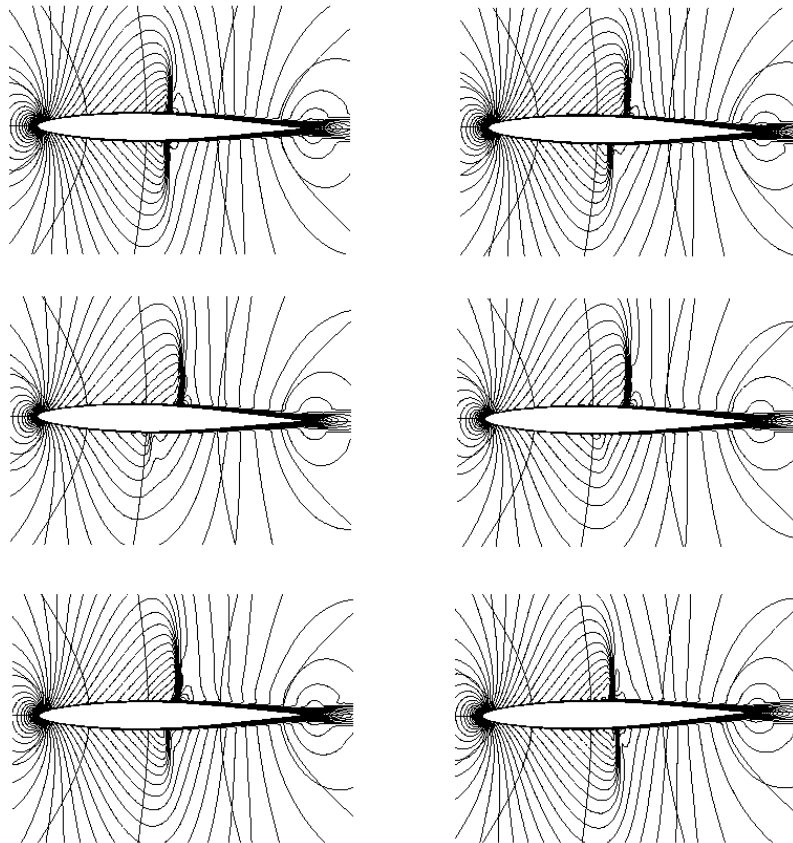
- **RK “dual time stepping”** scheme with variable local Δt^* (RK-BDF)
- Nonlinear **SGS “dual time stepping”** scheme (SGS-BDF)

with **Multigrid**

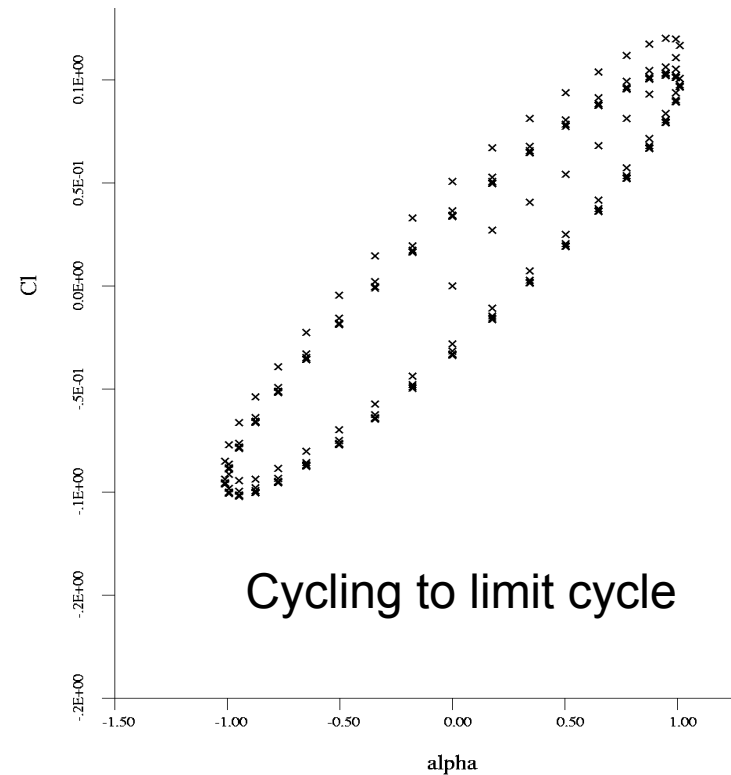


Test Case: NACA64A010 pitching airfoil (CT6 Case)

Mach Number	0.796
Pitching amplitude	+/- 1.01deg.
Reduced Freq.	0.202
Reynolds Number	12.36 million



Pressure Contours at Various Time Instances (AGARD 702)



Results of SGS-BDF Scheme

(36 time steps per pitching cycle,
3 iterations per time step)



Payoff of Dual-time Stepping BDF Schemes

- Accurate simulations with an **order of magnitude reduction** in time steps.
- For the pitching airfoil:
from ~ 1000 to 36 time steps per pitching cycle
with three sub-iterations in each step.



Frequency Domain and Global Space-Time Multigrid Spectral Methods

Application : Time-periodic flows

Using a **Fourier representation** in time, the time period T is divided into N steps.

Then,

$$\hat{w}_k = \frac{1}{N} \sum_{n=0}^{N-1} w^n e^{ikn\Delta t}$$

The discretization operator is given by

$$D_t w^n = \sum_{k=-\frac{N}{2}}^{\frac{N}{2}-1} ik\hat{w}_k e^{ikn\Delta t}$$



Method 1 (McMullen et.al.) : Transform the equations into **frequency domain** and solve them in pseudo-time t^*

$$\frac{d\hat{w}_k}{dt^*} + ik\hat{w}_k + \hat{R}_k = 0$$

Method 2 (Gopinath et.al.) : Solve the equations in the time-domain. The space-time spectral discretization operator is

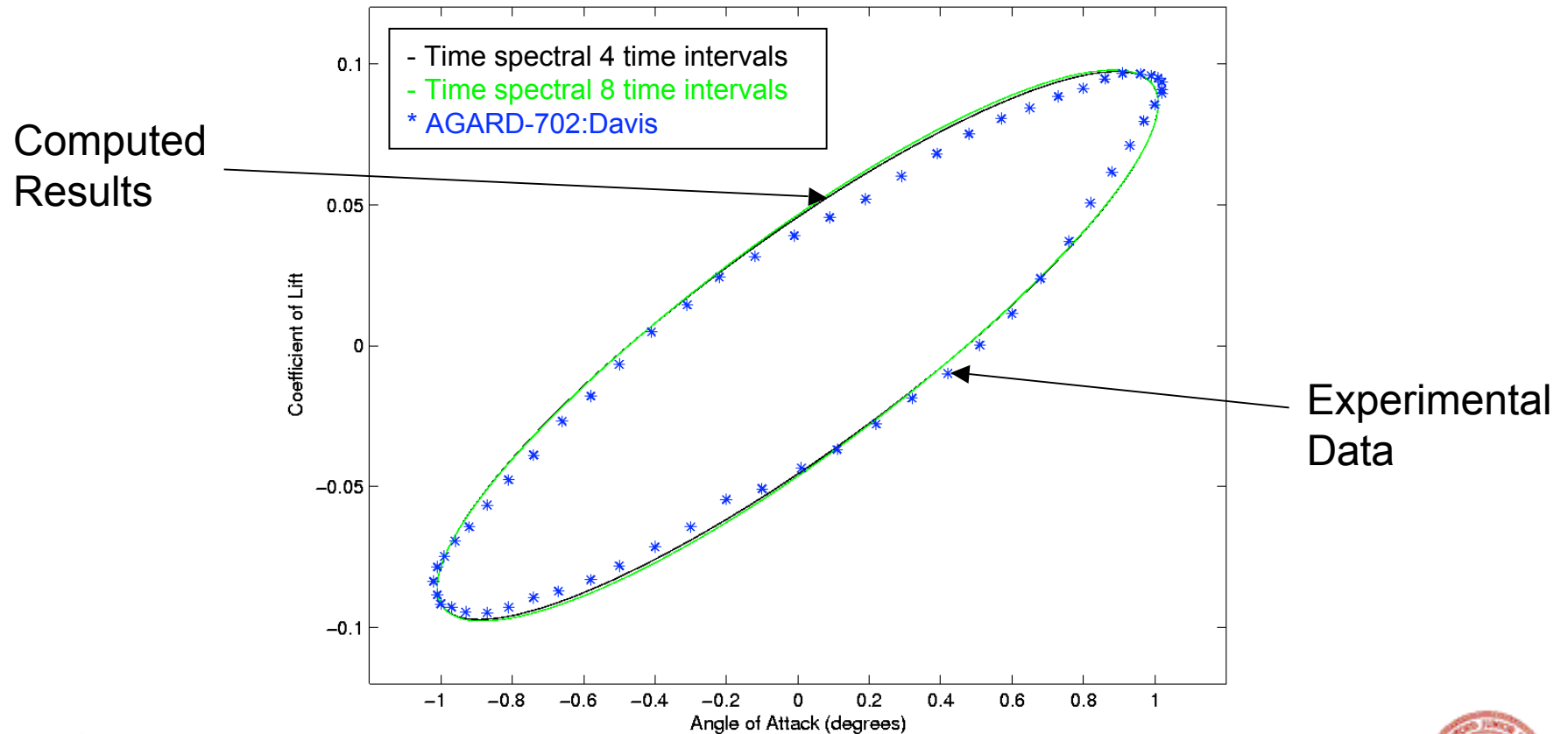
$$D_t w^n = \sum_{m=-\frac{N}{2}+1}^{\frac{N}{2}-1} d_m w^{n+m}, \quad d_m = \frac{1}{2} (\Delta t)^{m+1} \cot\left(\frac{\Delta t m}{N}\right), m \neq 0$$

This is a central difference operator connecting all time levels, yielding an **integrated space-time formulation** which requires simultaneous solution of the equations at all time levels.



Comparison with Experimental Data - C_L vs. α

RANS Time-Spectral Solution with 4 and 8 intervals per pitching cycle



Payoff of Time Spectral Schemes

- Engineering accuracy with very **small number of time intervals** and same rate of convergence as the BDF.
- **Spectral accuracy** for sufficiently smooth solutions.
- Periodic solutions directly without the need to evolve through 5-10 cycles, yielding **an order of magnitude reduction** in computing cost beyond the reduction already achieved with the BDF, for a total of **two orders of magnitude**.



CFD Algorithms: Open Issues

- Capability of turbulence models to predict **transition** and **separation**
- Prediction of transitional flows - which may be very important for **small UAVs**
- Reliable a posteriori **error bounds** based on the computed result
- Fast convergence of viscous solutions on **arbitrary meshes** (perhaps with the SGS multigrid scheme)
- BGK Scheme for RANS equations, hypersonic and rarefied gas flows



Proposal

- Collaborative development between Boeing, Intelligent Aerodynamic Inc. and Stanford University of advanced algorithms to be incorporated in industrial-strength software to provide:
assured accuracy, throughput, and turnaround.
- Funding, and distribution of intellectual property rights of derivative software, to be negotiated.



Part 3

Application of Shape Optimization via Control Theory

to Aerodynamic Design,
with Potential for Other Disciplines

Presented by Antony Jameson and
Kasidit Leoviriyakit



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Outline

- Automatic Design via Control Theory
- Planform and Aero-Structural Optimization
- Design using an Unstructured Mesh



Note on the History of the Adjoint Method for Transonic and Supersonic Wing Design

- Since the original proposal to apply the adjoint method to transonic wing design (Jameson, 1988), shape optimization via control theory has been the subject of 15 years intensive development.
- Multiple sources of funding, including Air Force Office of Scientific Research.
- First numerical results: *Jameson 1989, Science Vol 245, 361 - 371*
- Applications to Beech Premier, MDXX 1995 - 1996
- Application to HSCT 1997 - 2000



Levels of CFD

HIGHEST

Automatic Design

- Integrate the predictive capability into an automatic design method that incorporates computer optimization.

Interactive Calculation

- Attainable when flow calculation can be performed fast enough
- But does NOT provide any guidance on how to change the shape if performance is unsatisfactory.

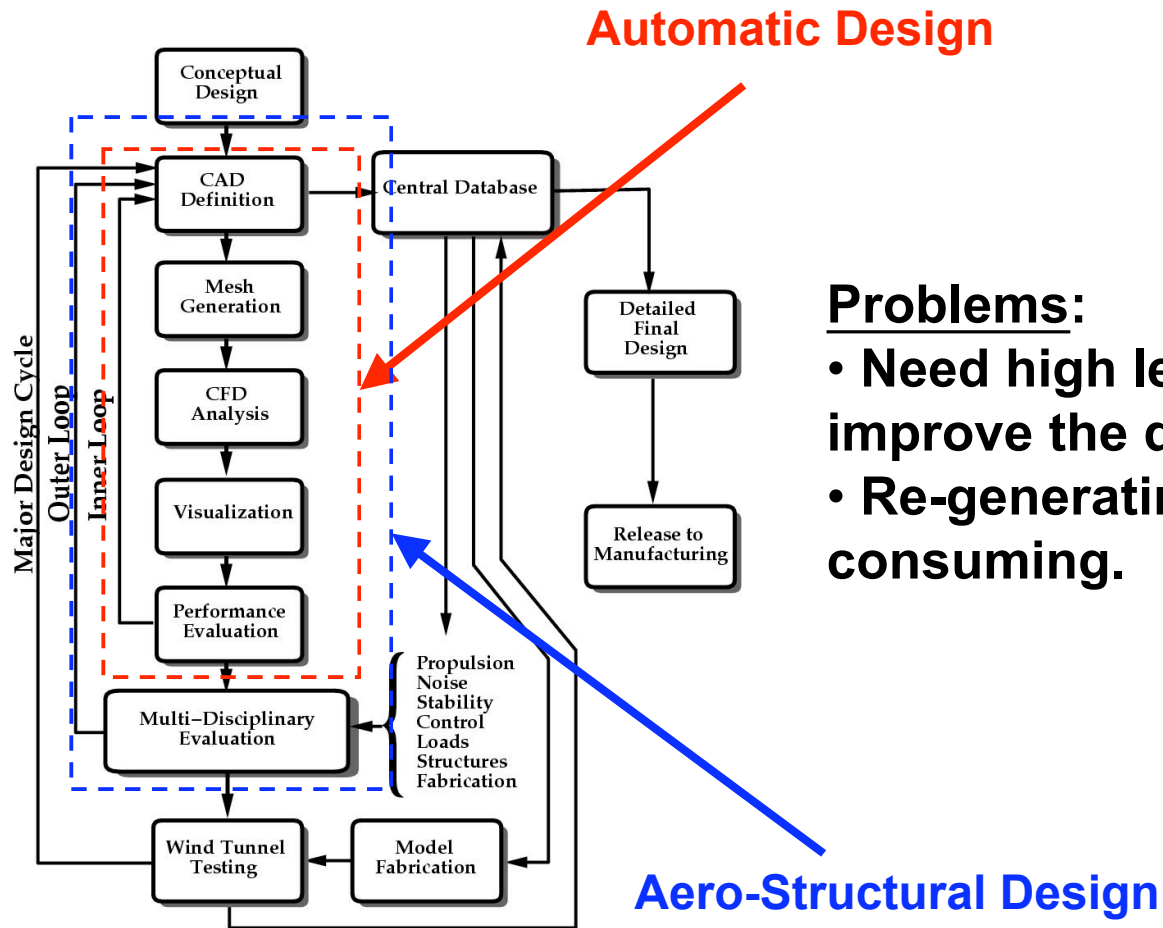
Flow Prediction

- Predict the flow past an airplane or its important components in different flight regimes such as take-off or cruise and off-design conditions such as flutter.
- Substantial progress has been made during the last decade.

LOWEST



Aerodynamic Design Process



Problems:

- Need high level of expertise to improve the design.
- Re-generating mesh is time consuming.

Preliminary Design Level



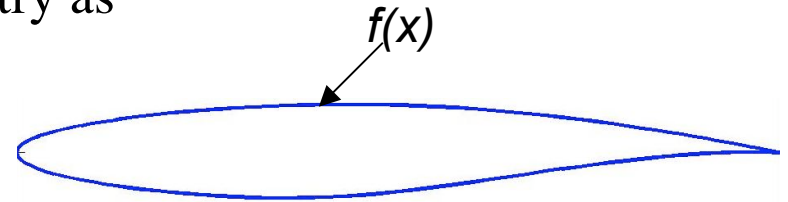
Optimization and Design using Sensitivities Calculated by the Finite Difference Method

The simplest approach is to define the geometry as

$$f(x) = \sum \omega_i b_i(x)$$

where ω_i = weight,

$b_i(x)$ = set of shape functions



Then using the finite difference method, a cost function

$$I = I(w, \omega) \quad (\text{such as } C_D \text{ at constant } C_L)$$

has sensitivities

$$\frac{\partial I}{\partial \omega_i} = \frac{I(\omega_i + \Delta \omega_i) - I(\omega_i)}{\Delta \omega_i}$$

If the shape changes is

$$\omega^{n+1} = \omega^n + \Delta \omega \frac{\partial I}{\partial \omega_i} \quad (\text{with small positive } \Delta \omega)$$

The resulting improvements is $I + \Delta I = I + \Delta \omega \frac{\partial I}{\partial \omega} = I + \Delta \omega \frac{\partial I}{\partial \omega} < I$

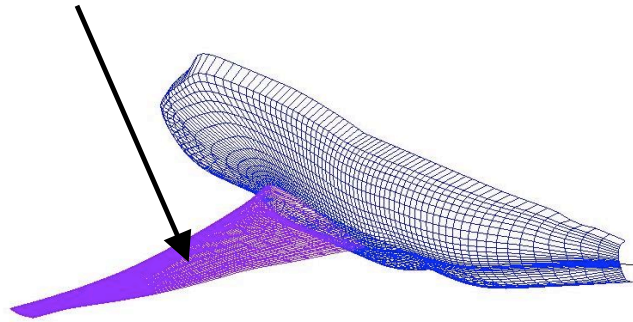
More sophisticated search may be used, such as quasi-Newton.



Disadvantage of the Finite Difference Method

The need for a number of flow calculations proportional to the number of design variables

Using 4224 mesh points
on the wing as design variables



Boeing 747

4225 flow calculations
~ 30 minutes each (RANS)

Too Expensive



Application of Control Theory

GOAL : Drastic Reduction of the Computational Costs

Drag Minimization \equiv **Optimal Control of Flow Equations**
subject to Shape(wing) Variations

Define the cost function

$$I = I(w, F) \quad (\text{for example } C_D \text{ at fixed } C_L)$$

and a change in F results in a change

$$\Delta I = \begin{bmatrix} \frac{\partial I}{\partial w} \\ \frac{\partial I}{\partial F} \end{bmatrix}^T \Delta w + \begin{bmatrix} \frac{\partial I}{\partial w} \\ \frac{\partial I}{\partial F} \end{bmatrix}^T \Delta F$$

Suppose that the governing equation R which expresses the dependence of w and F as

$$R(w, F) = 0$$

and

$$\Delta R = \begin{bmatrix} \frac{\partial R}{\partial w} \\ \frac{\partial R}{\partial F} \end{bmatrix}^T \Delta w + \begin{bmatrix} \frac{\partial R}{\partial w} \\ \frac{\partial R}{\partial F} \end{bmatrix}^T \Delta F = 0$$



Application of Control Theory

Since the variation δR is zero, it can be multiplied by a Lagrange Multiplier λ and subtracted from the variation δI without changing the result.

$$\begin{aligned} \delta I &= \frac{\partial I^T}{\partial w} \delta w + \frac{\partial I^T}{\partial F} \delta F - \lambda^T \left(\frac{\partial R}{\partial w} \delta w + \frac{\partial R}{\partial F} \delta F \right) \\ &= \left(\frac{\partial I^T}{\partial w} - \lambda^T \frac{\partial R}{\partial w} \right) \delta w + \left(\frac{\partial I^T}{\partial F} - \lambda^T \frac{\partial R}{\partial F} \right) \delta F \end{aligned}$$

Choosing λ to satisfy the adjoint equation

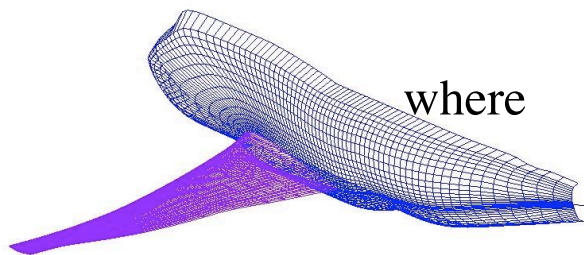
$$\lambda^T \frac{\partial R}{\partial w} = \frac{\partial I^T}{\partial w} \quad (\text{Adjoint Equation})$$

the first term is eliminated, and we find that

$$\delta I = G^T \delta F$$

where

$$G^T = \frac{\partial I^T}{\partial F} - \lambda^T \frac{\partial R}{\partial F} \quad (\text{Gradient})$$



4224 design variables

One Flow Solution + One Adjoint Solution

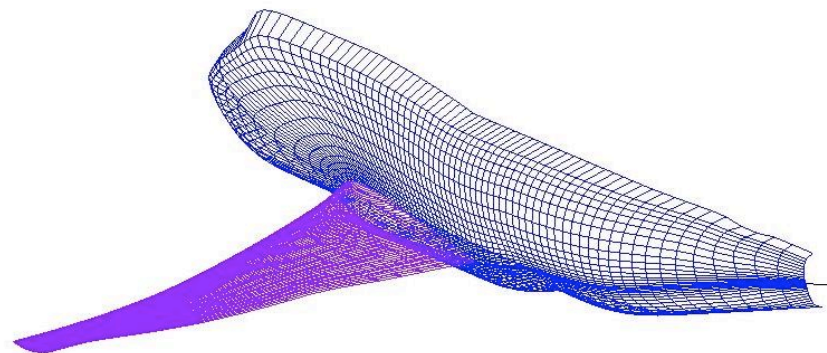


Advantages of the Adjoint Method:

- Gradient for N design variables with cost equivalent to two flow solutions

- Minimal memory requirement in comparison with automatic differentiation

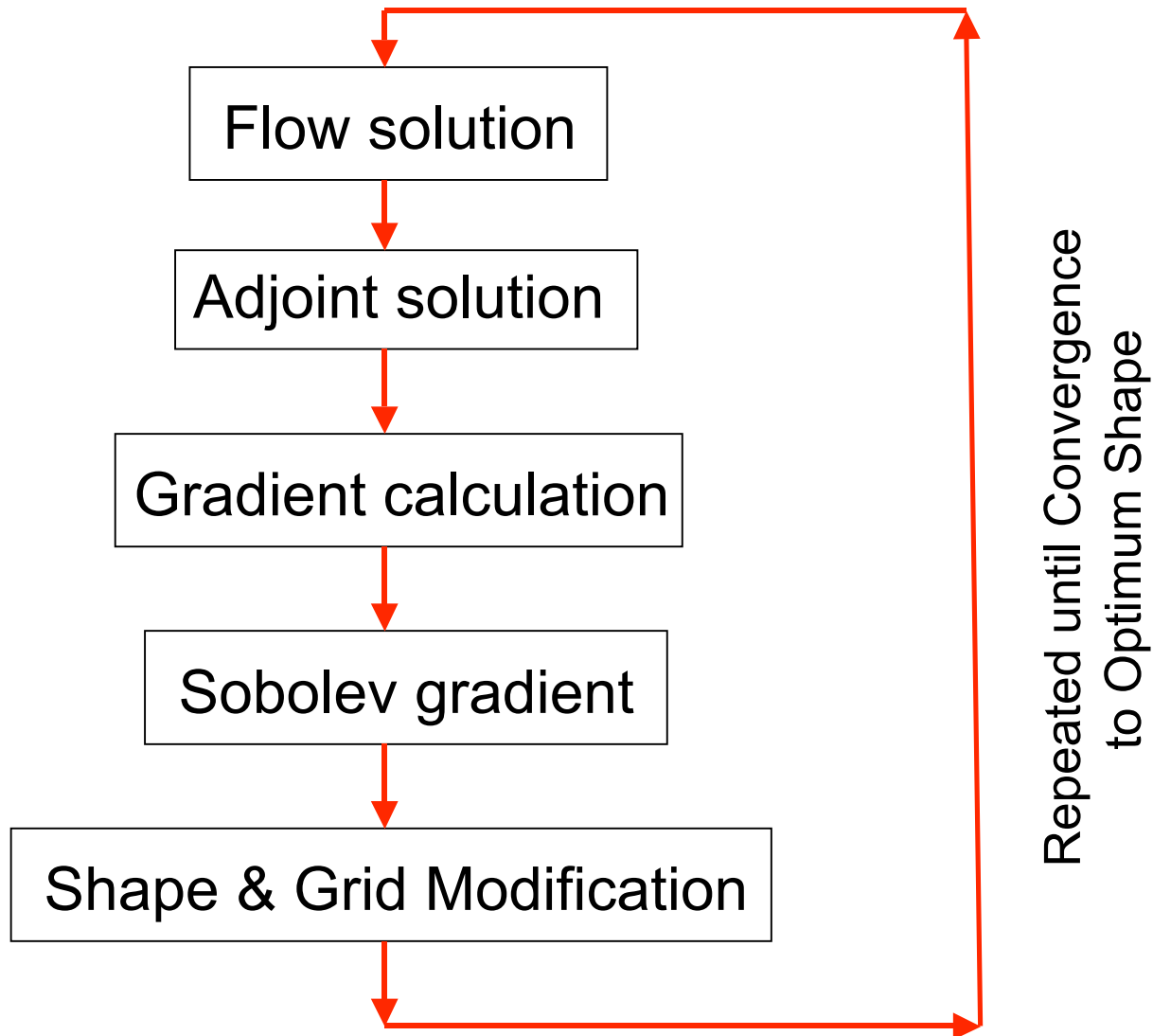
- Enables shapes to be designed as free surface
 - No need for user defined shape function
 - No restriction on the design space



4224 design variables



Outline of the Design Process



Discrete versus Continuous Adjoint Methods

- The discrete adjoint method evaluates the adjoint and gradient equations algebraically from the discretized flow equations.
- The continuous adjoint method evaluates the costate solution from the partial differential adjoint equation.
- The continuous adjoint method leads to no inconsistency as long as it is combined with a compatible search method
- In the limit of grid convergence the two approaches yield identical gradients.
- Numerical tests of a model problem verify slightly superior accuracy with the continuous formulation
(*Jameson and Vassberg 2000*)



Summary of the Continuous Flow and Adjoint Equations

With computational coordinates ξ_i

Euler equations for the flow :

$$(1) \quad \frac{\partial}{\partial \xi_i} S_{ij} f_j(w) = 0$$

where S_{ij} are metrics, $f_j(w)$ the fluxes.

Adjoint equation

$$(2) \quad C_i \frac{\partial \xi}{\partial \xi_i} = 0, \quad C_i = S_{ij} \frac{\partial f_j}{\partial w}$$

Boundary condition for the Inverse problem

$$(3) \quad I = \frac{1}{2} \int (p - p_t)^2 ds$$

$$\xi_2 n_x + \xi_3 n_y + \xi_3 n_z = p - p_t$$

Gradient

$$(4) \quad \frac{\partial I}{\partial w} = \int \frac{\partial \xi}{\partial w} \left[S_{ij} f_j dD - \int_{\xi_w} (\xi_{21} \xi_2 + \xi_{22} \xi_3 + \xi_{23} \xi_4) p d\xi_1 d\xi_3 \right]$$



Sobolev Gradient

Key issue for successful implementation of the Continuous adjoint method.

Define the gradient with respect to the Sobolev inner product

$$\langle f, g \rangle = \int (\bar{g}f + \bar{g}'f') dx$$

Set

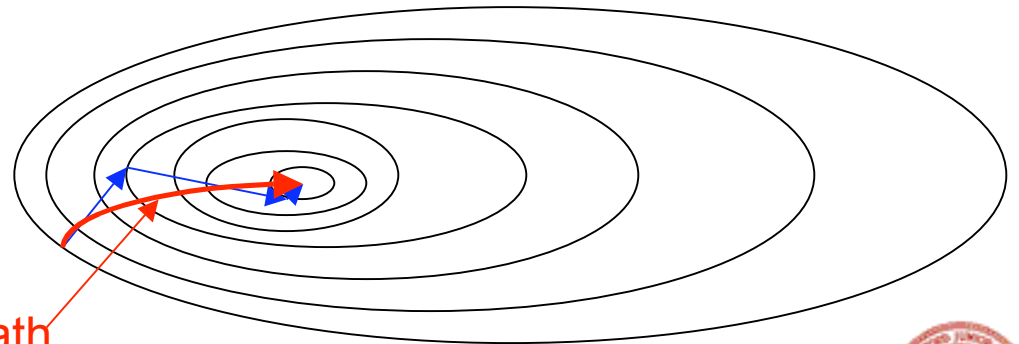
$$\bar{f} = \int \bar{g}, \quad \bar{f}' = \int \bar{g}'$$

This approximates a continuous descent process

$$\frac{df}{dt} = \bar{g}$$

The Sobolev gradient \bar{g} is obtained from the simple gradient g by the smoothing equation

$$\bar{g} \left[\frac{\partial}{\partial x} \left[\frac{\partial \bar{g}}{\partial x} \right] \right] = g.$$



Continuous descent path



Computational Costs with N Design Variables

(Jameson and Vassberg 2000)

Cost of Search Algorithm

Steepest Descent	$\mathcal{O}(N^2)$
Quasi-Newton	$\mathcal{O}(N)$
Sobolev Gradient	$\mathcal{O}(K)$

(Note: K is independent of N)

Total Computational Cost of Design

Finite Difference Gradients + Steepest Descent	$\mathcal{O}(N^3)$
Finite Difference Gradients + Quasi-Newton Search or Response surface	$\mathcal{O}(N^2)$
Adjoint Gradient + Quasi-Newton Search	$\mathcal{O}(N)$
Adjoint Gradient + Sobolev Gradient	$\mathcal{O}(K)$

(Note: K is independent of N)

- $N \sim 2000$
- Big Savings
- Enables Calculations on a Laptop



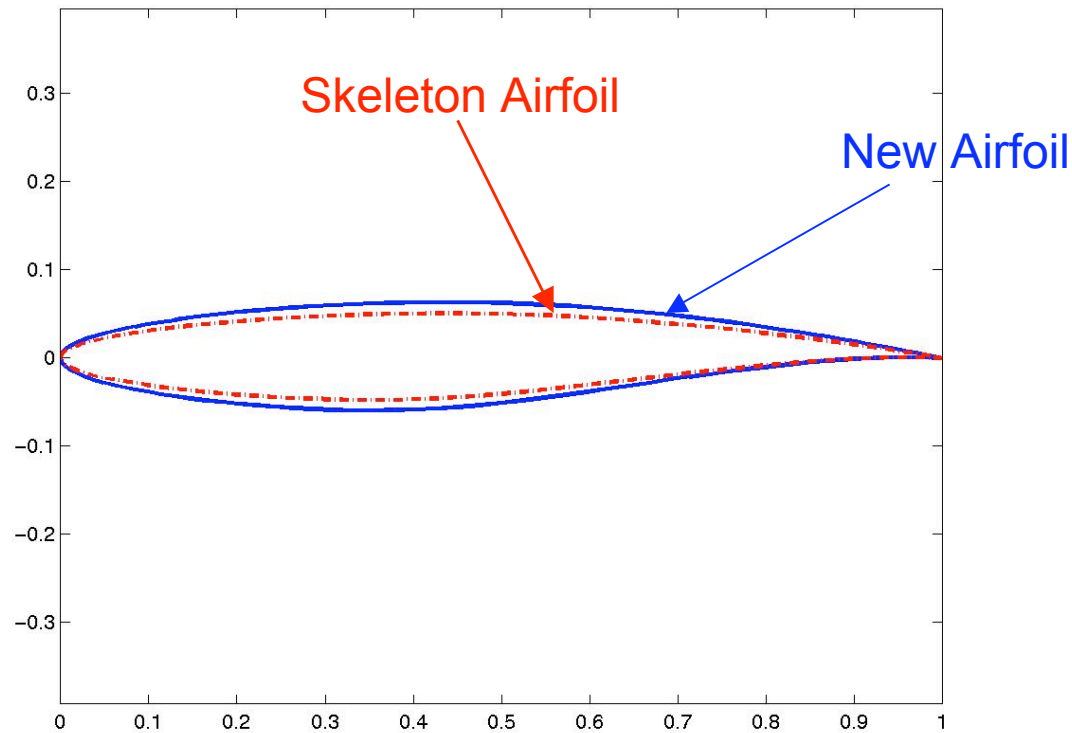
Constraints Enforced in SYN107

For drag minimization

1. Fixed C_L
2. Fixed span load
 - Keep out-board C_L low enough to prevent buffet
 - Fixed root bending moment
3. Maintain specified thickness
 - Sustain root bending moment with equal structure weight
 - Maintain fuel volume
4. Smooth curvature variations via Sobolev gradient



Application of Thickness Constraint



- Prevent shape change penetrating a specified skeleton
- Separate thickness and camber to allow free camber variations
- Minimal user input required.



Multi-Point Design

- Drag minimization at a **single design** point typically produces a **shock free** flow but may **adversely affect performance** at **other points**. (Double shocks below design point, drag creep due to leading edge shock)
- Good **all round performance** can be enhanced by **multi-point design**
e.g. Mach = 0.75 $C_L = 0.5$ (drag creep)
 Mach = 0.86 $C_L = 0.42$ (cruise)
 Mach = 0.89 $C_L = 0.40$ (drag divergence)

Partial Redesign with Structural Constraints

Fixed (Structure box)



- **Design changes** can be limited to a specified spanwise range of the wing
- **Section changes** can be limited to a specified chordwise range
- The **shape changes** are blended smoothly via the **Sobolev gradient**



Inverse Design

- A hard test :
 - ONERA M6 Wing target pressure
Mach 0.84
□ 3.06 degree
Lambda Shock
 - Starting from NACA0012 sections
(single shock)
- Recovery of smooth symmetric profile
from discontinuous lifting pressure distribution

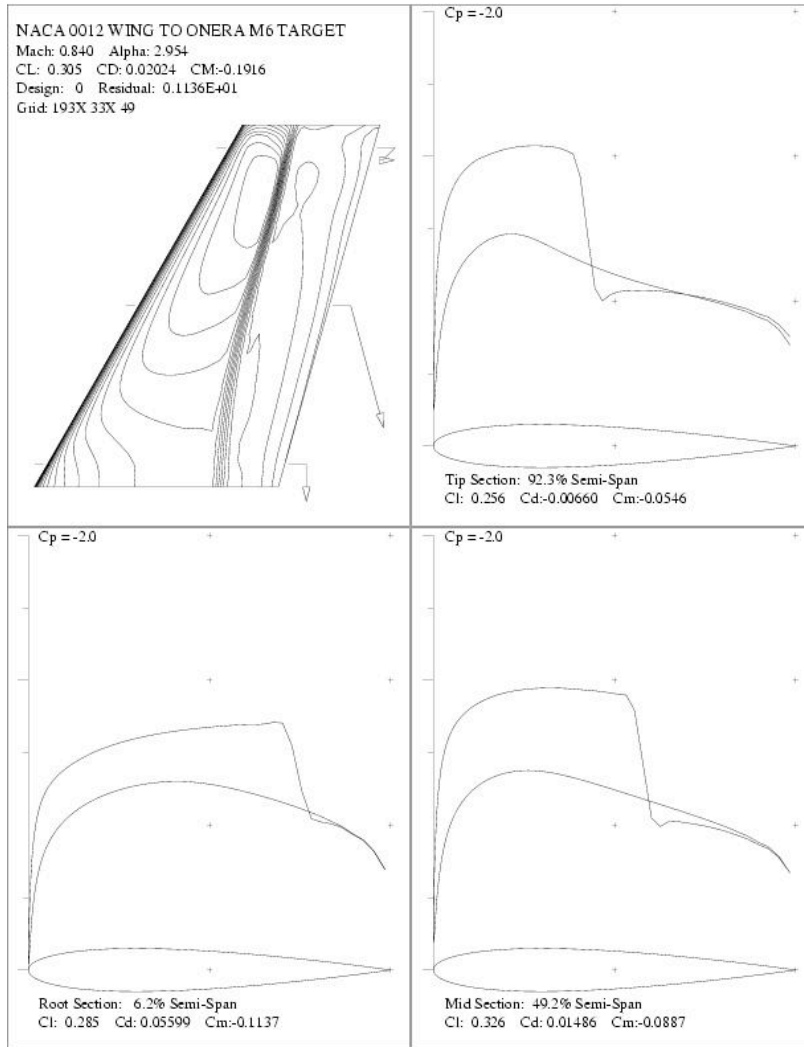


Inverse Design

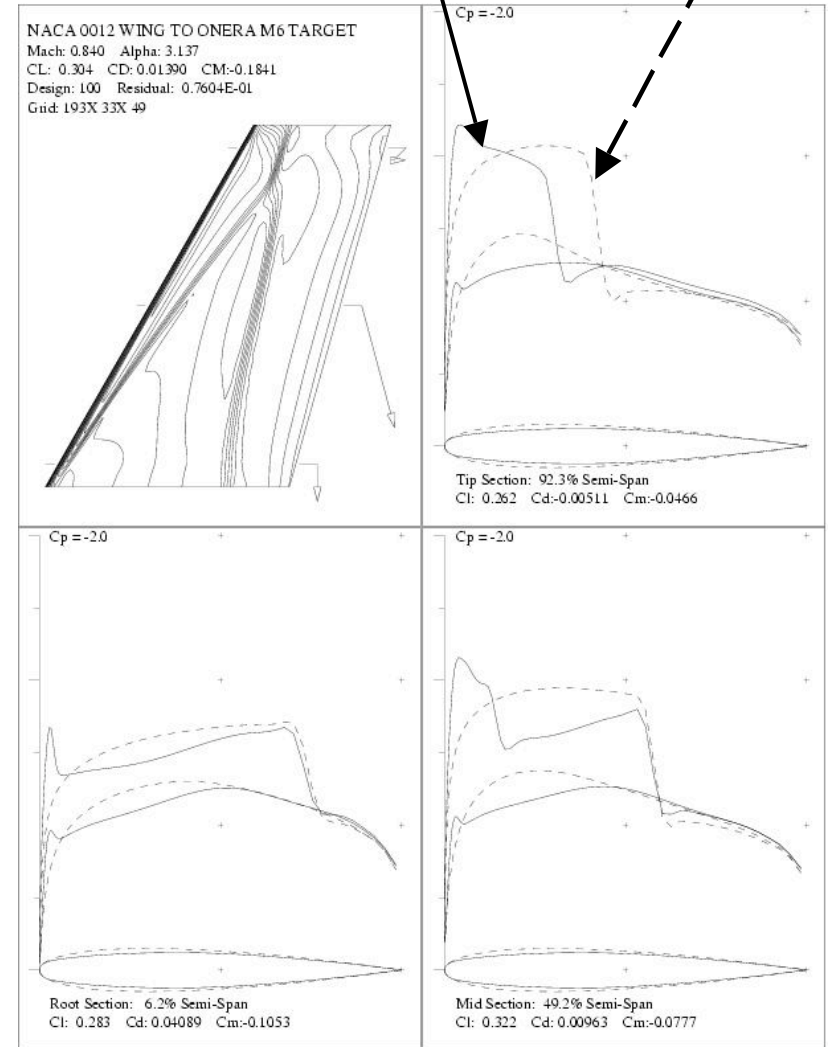
Start with NACA 0012

Target

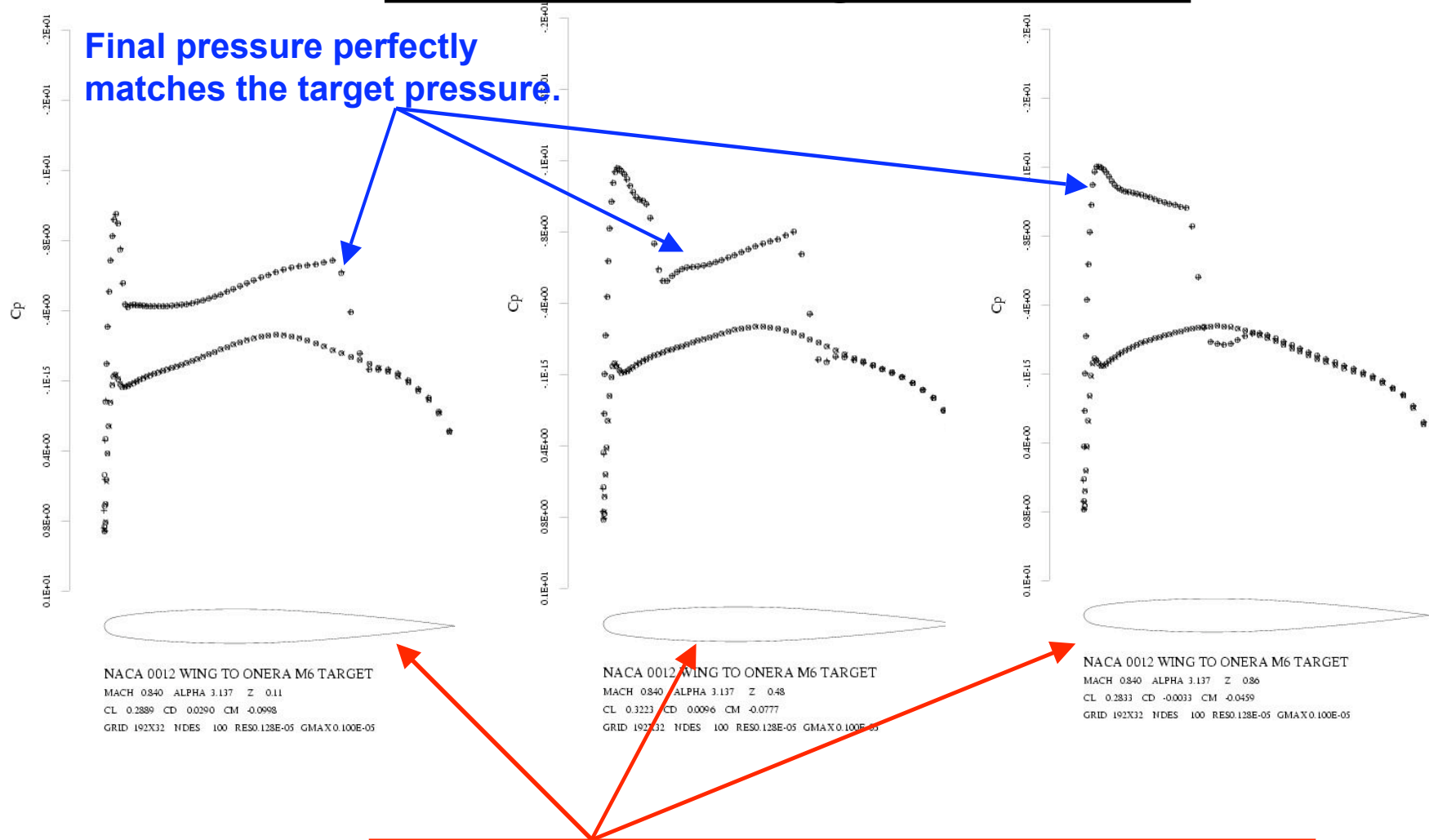
Initial



Matching ONERA M6 Target Pressure



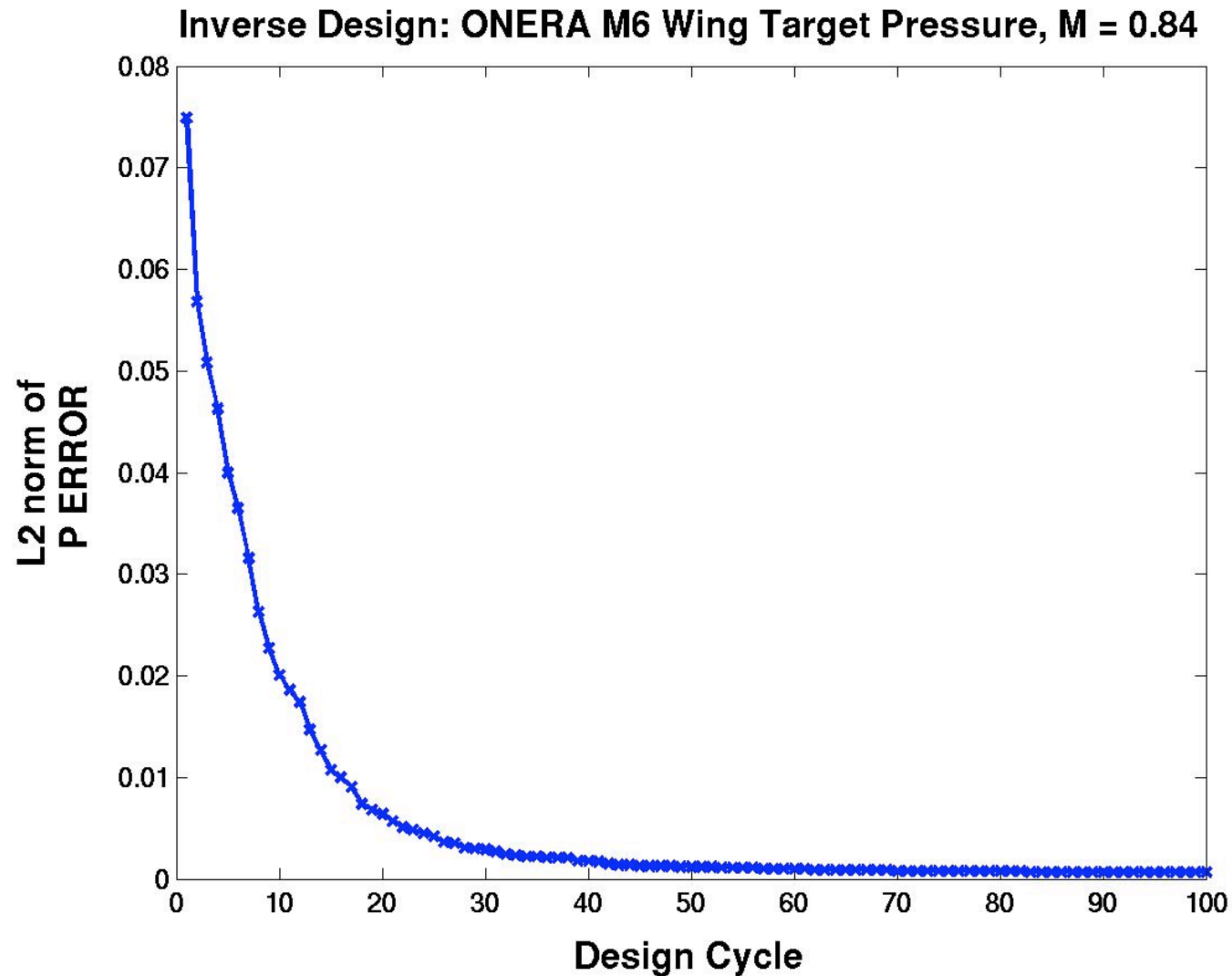
Inverse Design Results



Recover ONERA M6 Shape at all span stations



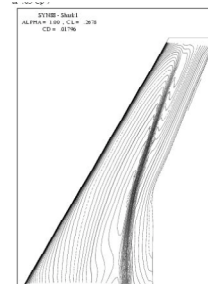
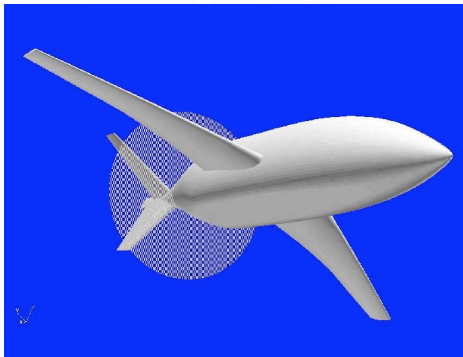
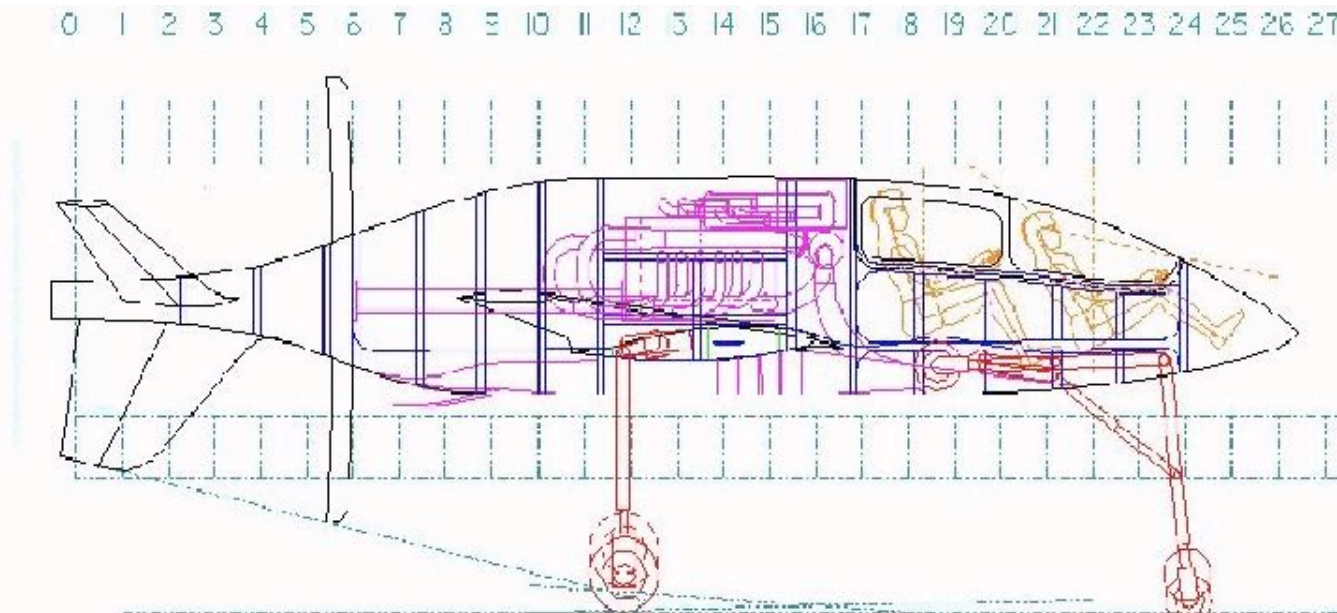
Convergence History of Inverse Design



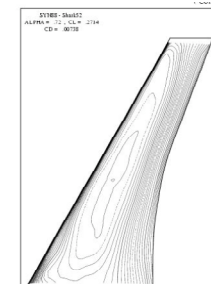
Applications to Unconventional Designs

Shark Reno Air Racer

(Ahlstrom, Gregg, Vassberg, Jameson, AIAA PAPER 2000-4341)



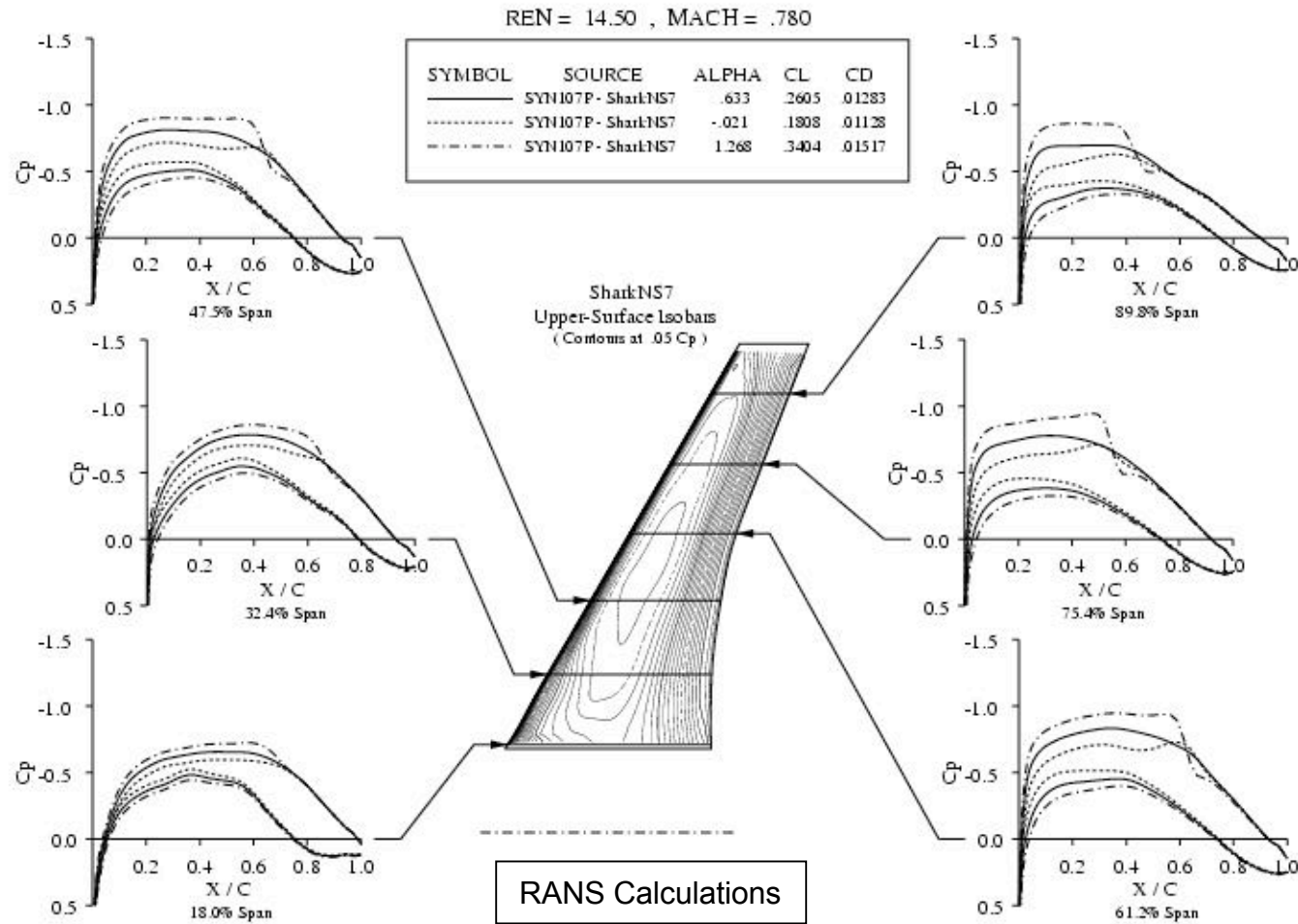
Original



Redesign

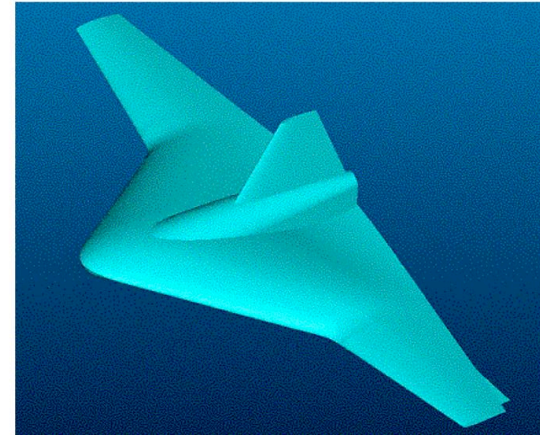
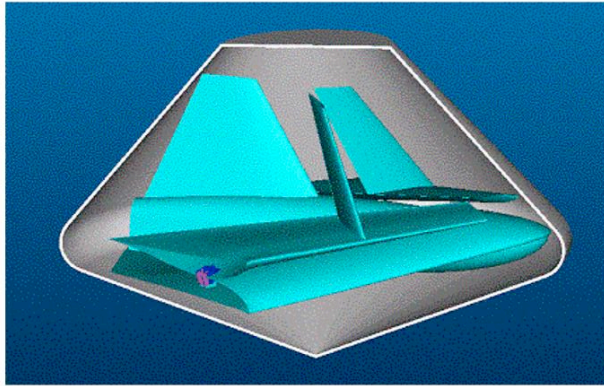


Shark Reno Air Racer



Mars Lander

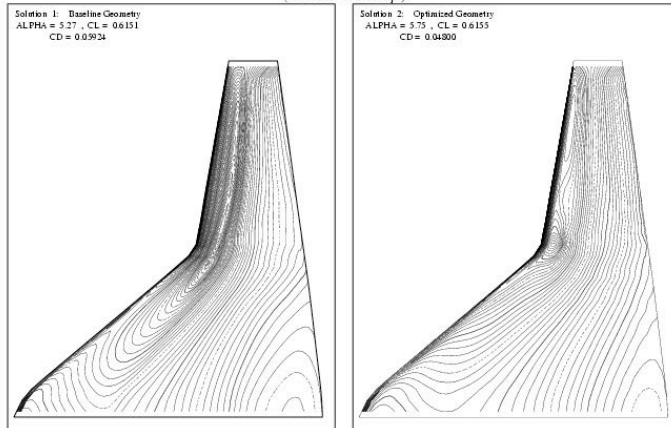
(Vassberg, Jameson 2004)



MARES Wing Design - Drag Polars
SYN107P Drag Minimization

MARES Wing Design - Upper Surface Isobars
SYN107P Drag Minimization
REN = 170K, MACH = 0.650
(Contours at 0.05 Cp)

RANS Calculations



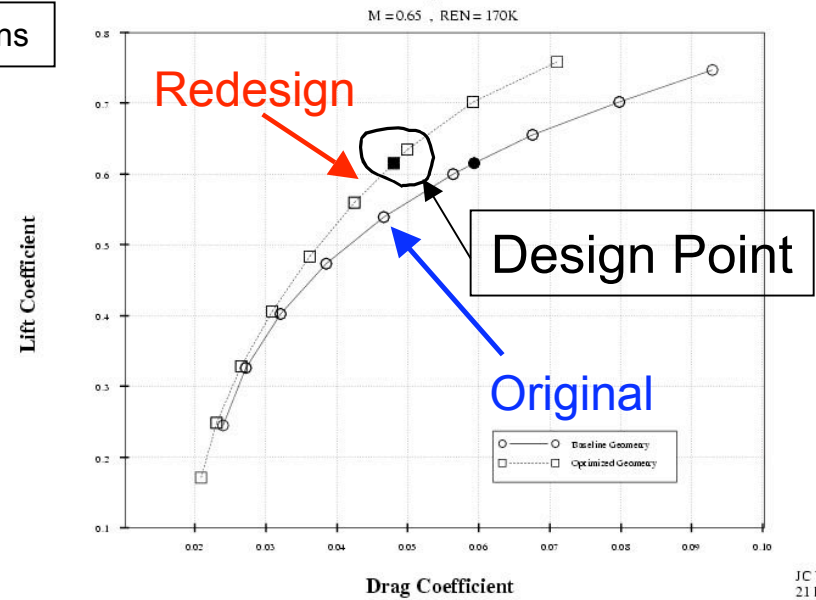
COMPPLOT
Ver 2.00



John C. Vassberg
12:07 Sat
20 Dec 03

Original

Redesign



JC Vassberg
21 Dec 2003



Drag reduced by 20 %



Planform and Aero-Structural Optimization

(Leoviriyakit, Jameson 2003 - 2004)

- Design tradeoffs suggest an multi-disciplinary design and optimization

$$\text{Range} = \frac{VL}{D} \frac{1}{\text{sfc}} \log \frac{W_o + W_f}{W_o}$$

Maximize Minimize



Planform variations can further maximize VL/D but affects W_o



Aerodynamic Design Tradeoffs

The drag coefficient can be split into

$$C_D = C_{DO} + \frac{C_L^2}{\pi e AR}$$

$\frac{L}{D}$ is maximized if the two terms are equal.

Induced drag is half of the total drag.

If we want to have large drag reduction, we should target the induced drag.

$$D_i = \frac{2L^2}{\pi e V^2 b^2}$$

Change span by changing planform

Design dilemma

Increase b

D_i decreases

W_o increases



Break Down of Drag

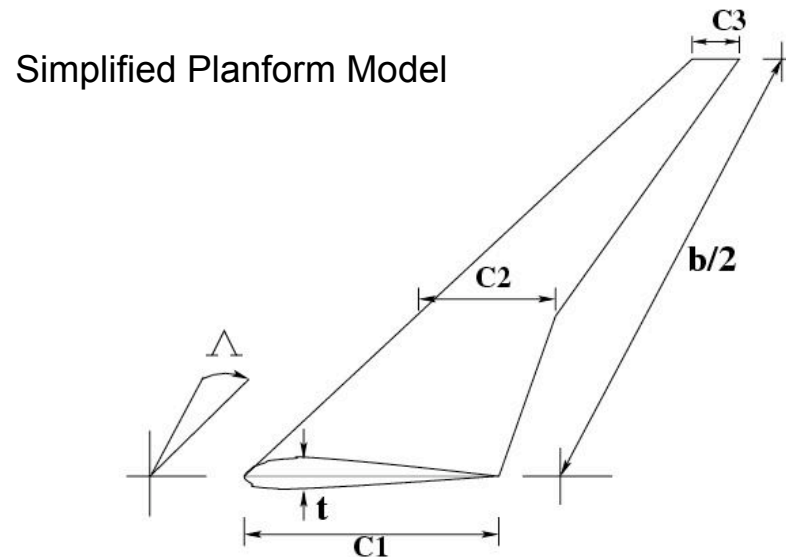
Boeing 747 at $C_L \sim .47$ (including fuselage lift $\sim 15\%$)

Item	C_D	Cumulative C_D
Wing Pressure	120 counts (15 shock, 105 induced)	120 counts
Wing friction	45	165
Fuselage	50	215
Tail	20	235
Nacelles	20	255
Other	15	270
Total	<u>270</u>	

Induced Drag is the largest component



Wing Planform Optimization



Wing planform modification can yield larger improvements **BUT** affects structural weight.

$$I = \alpha_1 C_D + \alpha_2 \frac{1}{2} \int (p - p_d)^2 dS + \alpha_3 C_W$$

where

$$C_W = \frac{\text{Structural Weight}}{q S_{ref}}$$

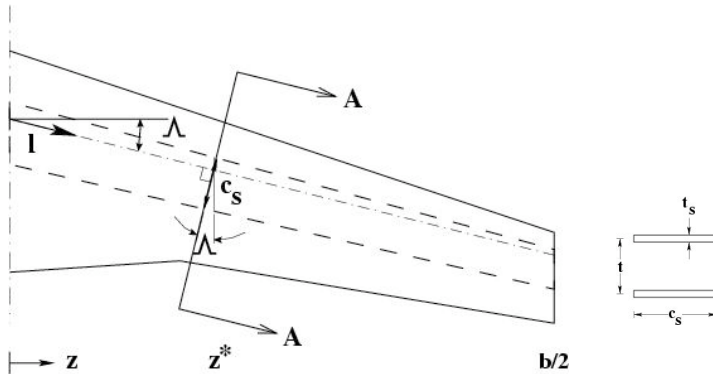
Can be thought of as constraints



Additional Features Needed

- Structural Weight Estimation
- Large scale gradient : span, sweep, etc...
- Adjoint gradient formulation for dC_w/dx
- Choice of \square_1 , \square_2 , and \square_3

Use wing box to estimate the structural weight.



Large scale gradient

- Use summation of mapped gradients to be large scale gradient



Choice of Weighting Constants

Breguet range equation

$$R = \frac{VL}{D} \frac{1}{sfc} \log \frac{W_o + W_f}{W_o}$$

With fixed V , L , sfc , and $(W_o + W_f \equiv W_{TO})$, the variation of R can be stated as

$$\frac{\Delta R}{R} = \frac{\Delta C_D}{C_D} + \frac{1}{\log \frac{W_{TO}}{W_o}} \frac{\Delta W_o}{W_o} = \frac{\Delta C_D}{C_D} + \frac{1}{\log \frac{C_{W_{TO}}}{C_{W_o}}} \frac{\Delta C_{W_o}}{C_{W_o}}$$

Maximizing Range \equiv **Minimizing**

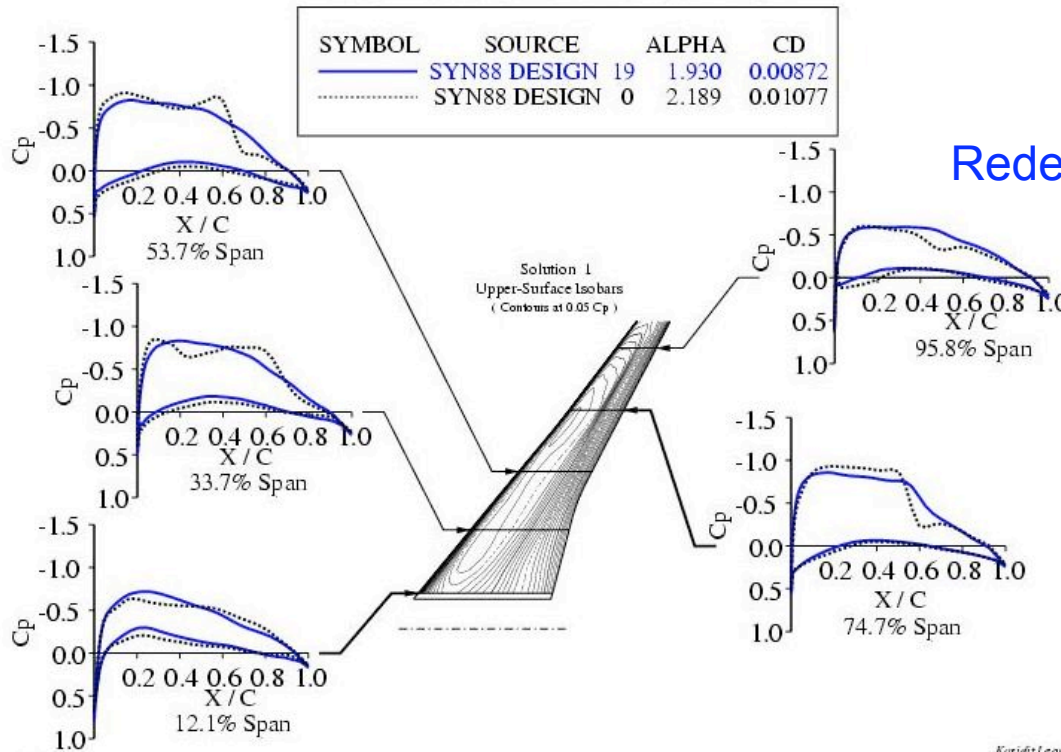
$I = C_D + \frac{\Delta_3}{\Delta_1} C_W$ using

$$\frac{\Delta_3}{\Delta_1} = \frac{C_D}{C_{W_o} \log \frac{C_{W_{TO}}}{C_{W_o}}}$$



Planform Optimization of Boeing 747

Constraints : Fixed $CL=0.42$



- 1) Longer span reduces the induced drag
- 2) Less sweep and thicker wing sections reduces structure weight
- 3) Section modification keeps shock drag minimum

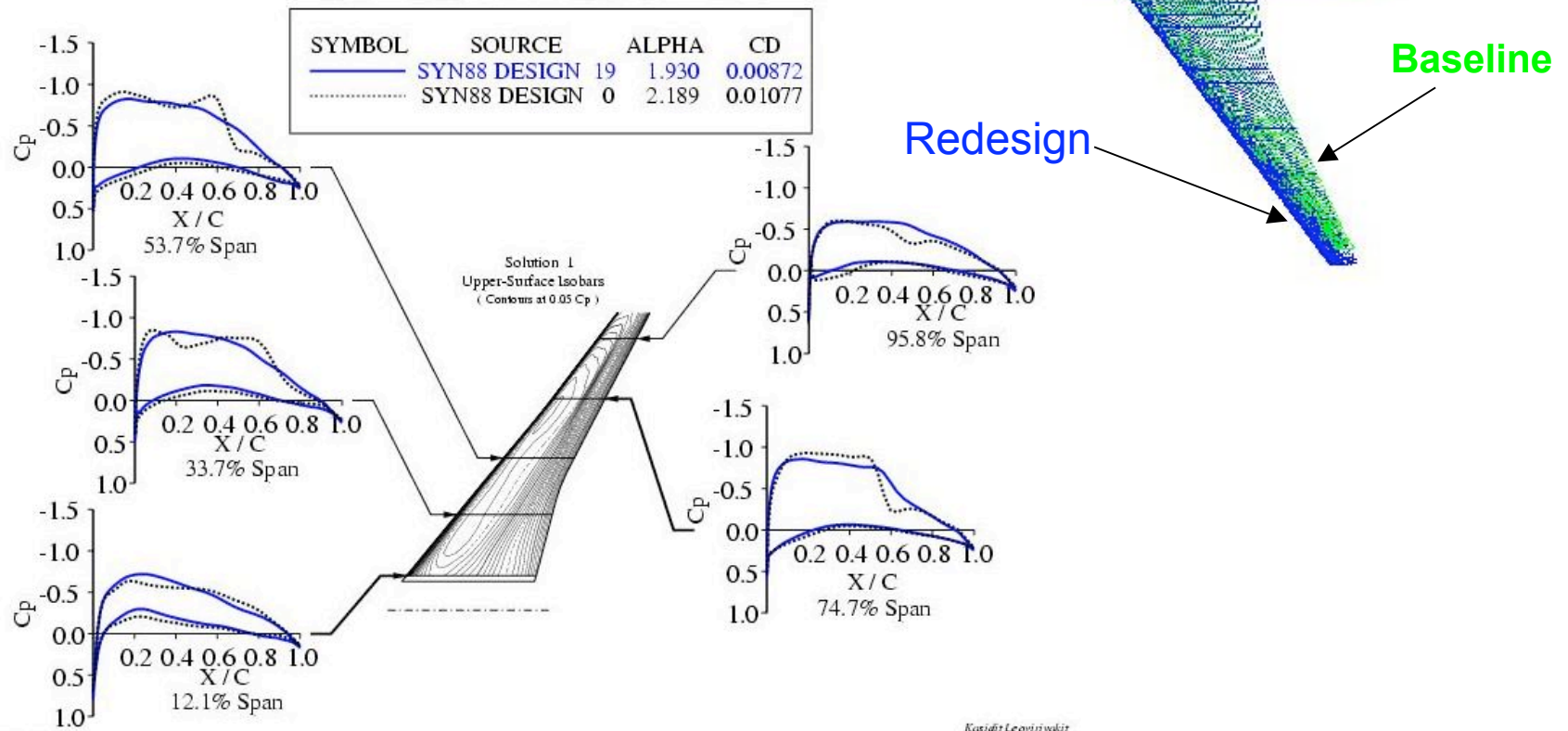
Overall: Drag and Weight Savings

	C_D	C_W
Baseline	108	455
Optimize Section at Fixed planform	94	455
Optimize both section and planform	87	450



Planform Optimization of MD11

Constraints : Fixed $CL=0.45$

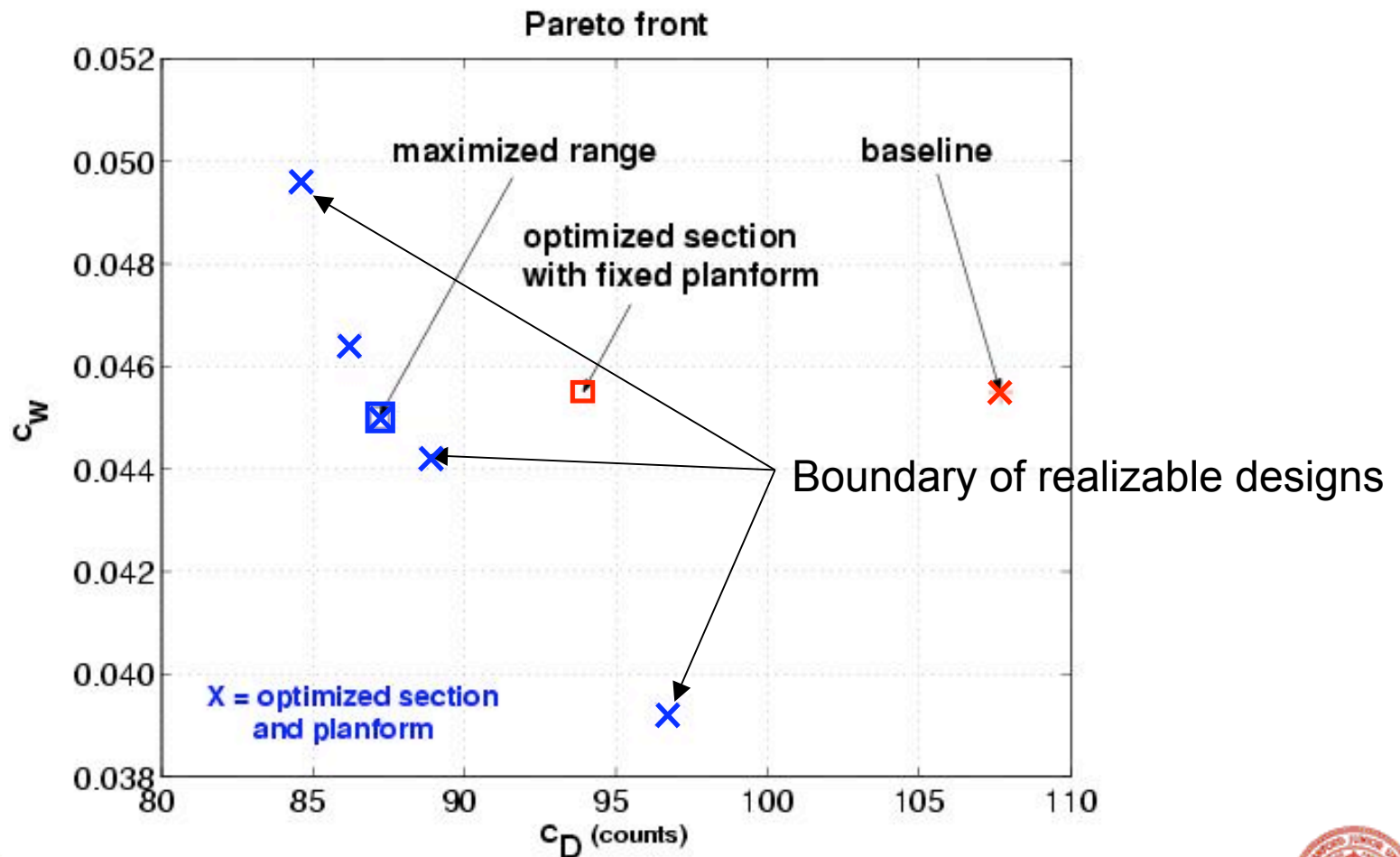


	C_D	C_W
Baseline	159	345
Optimize Section at Fixed planform	145	346
Optimize both section and planform	138	344



Pareto Front: “Expanding the Range of Designs”

Use multiple $\alpha_3/\alpha_1 \implies$ Multiple Optimal Shapes



Automatic design for the Complete Aircraft Geometry on an Unstructured Mesh (SYNPLANE)

- **Key step:** reduce the **gradient** to a **surface integral** independent of the mesh perturbation

(Jameson, A., and Kim, S., "Reduction of the Adjoint Gradient Formula in the Continuous Limit", 41 st AIAA Aerospace Sciences Meeting & Exhibit, AIAA Paper 2003-0040, Reno, NV, January 6-9, 2003.)

$$I = \int_{\Omega_w} \phi^T (\mathcal{S}_{2,j} f_j + C_2 w^*) d\Omega_1 d\Omega_3 \quad \int_{\Omega_w} (\mathcal{S}_{21} \Omega_2 + \mathcal{S}_{22} \Omega_3 + \mathcal{S}_{23} \Omega_4) p d\Omega_1 d\Omega_3$$

Compared to the previous formulation

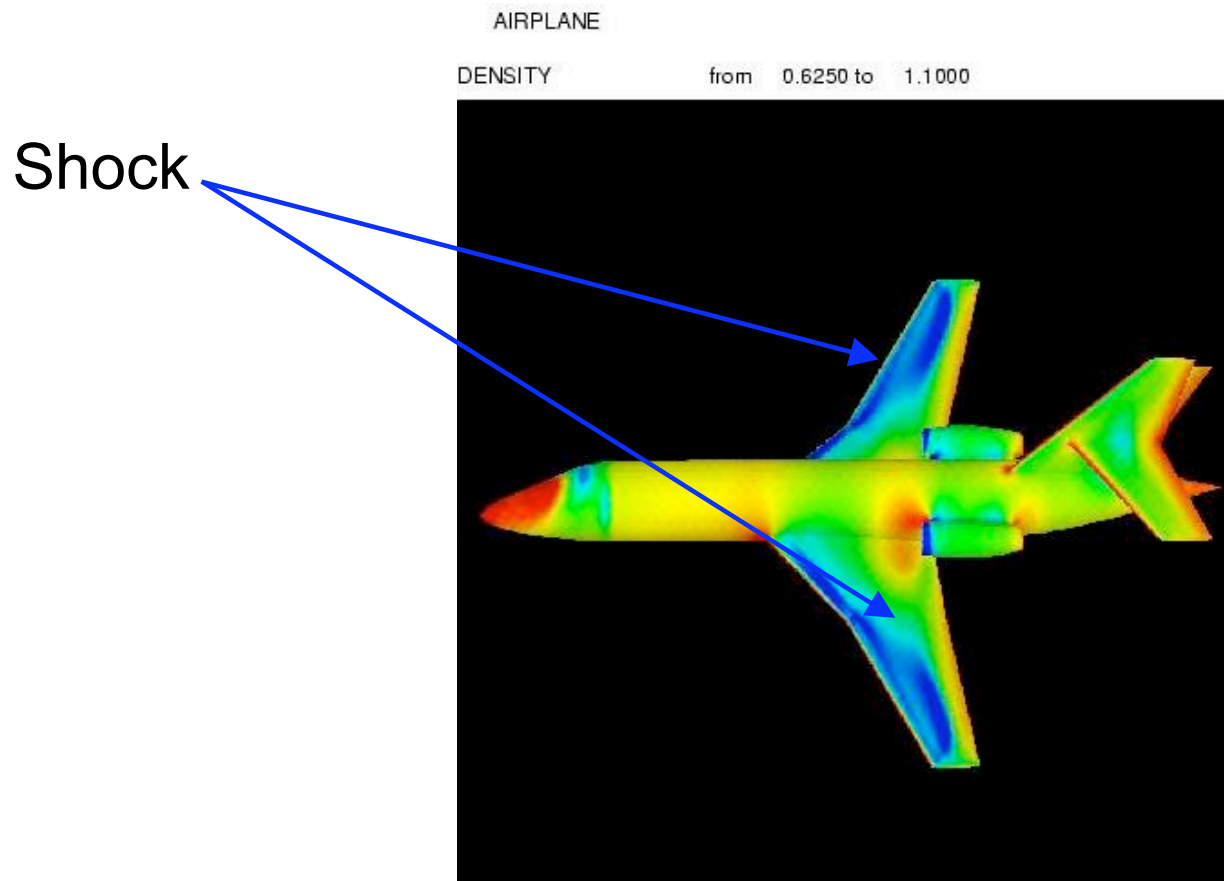
$$I = \int_D \frac{\partial \phi^T}{\partial \Omega_i} \mathcal{S}_{ij} f_j dD \quad \int_{\Omega_w} (\mathcal{S}_{21} \Omega_2 + \mathcal{S}_{22} \Omega_3 + \mathcal{S}_{23} \Omega_4) p d\Omega_1 d\Omega_3$$

This field integral is converted to boundary integral



Redesign of Falcon

Complete aircraft calculation on Unstructured Mesh



$C_D = 234$ counts



Redesign of Falcon

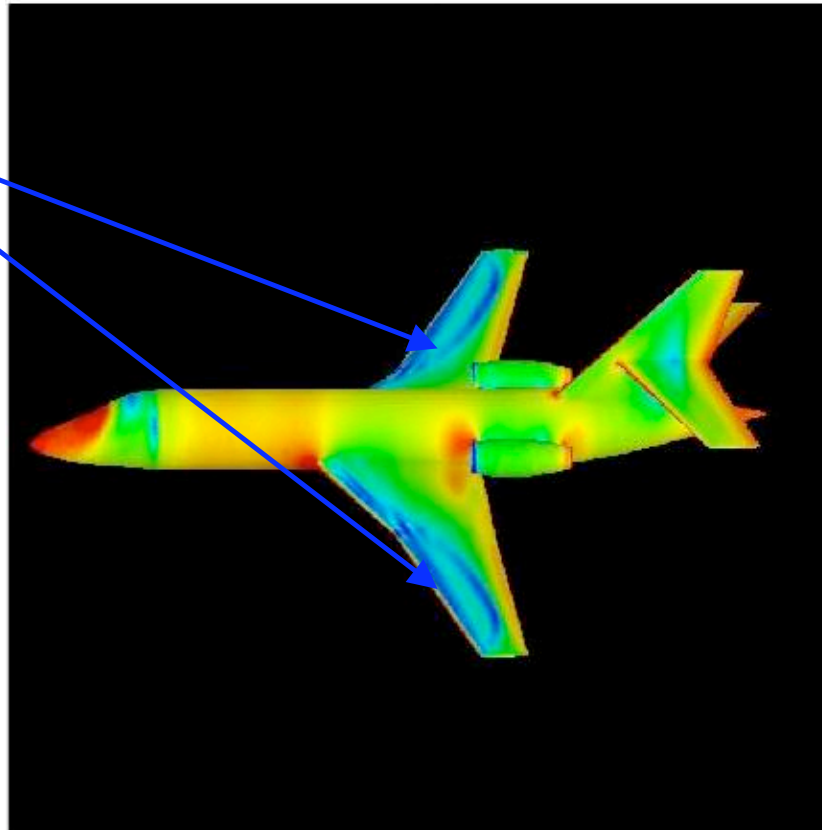
Using SYNPLANE

Drag reduction 18 counts at fixed $C_L = 0.4$

AIRPLANE

DENSITY from 0.6250 to 1.1000

Weakened Shock



$C_D = 216$ counts



Aerodynamic Shape Optimization Payoffs

- Enables aerodynamic design by **a small team** of experts focusing on the **true design issues** (e.g. Reno Air Racer)
- Significant reduction in **time** and **cost**
- Potential for **superior** and **unconventional** designs
- Potential for **“mid-course correction”** during the development cycle



Other Potential Applications of Shape Optimization

- Combined **shape** and **trajectory optimization** for **reentry vehicles** to reduce thermal loads
- Reduction of **acoustic signature**
(Take-Off and Landing noise, sonic boom reduction)
- Minimization of **electro magnetic signature** while meeting aerodynamic requirement
- Minimization of **wave resistance** of **ship hulls** (in process)



Other Applications of the Adjoint Method in CFD

- Automatic error estimation with
computed error bounds

(Barth and Deconinck, Lecture Notes in Computational Science and Engineering, Vol 25, Springer, 2002

Giles and Pierce, SIAM Review, Vol 42, 2000, 247 - 64)

- Automatic mesh adaptation based on error estimation
(Vendite and Darmofal, AIAA Paper 2003-3845)



Shape Optimization: Some Open Issues

- Embedding shape optimization in overall system optimization
- Multi-Disciplinary Optimization beyond aero-structural optimization
- Incomplete development for arbitrary grids
- Incomplete development for propulsion integration



Users of Intelligent Aerodynamics Design Software (SYN88, SYN107, SYNPLANE)

- *Raytheon Beach*
- *McDonnell Douglas*
- *Lockheed Martin Skunk Works*
- *Airbus UK*
- *SAAB*
- *IPTN*
- *Embraer (via NLR)*
- *Gulfstream*
- *Bombardier Aerospace (in contract)*
- *NASA Ames (HSCT, RLV)*

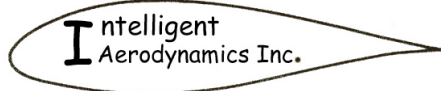


Proposal

- Collaborative development between Boeing, Intelligent Aerodynamics Inc., and Stanford University of the adjoint method to meet industrial requirements for a streamlined design process.
- Funding and distribution of intellectual property rights to be negotiated.
- Intelligent Aerodynamics has existing IP rights.
- Derivative codes might be co-owned and marketed.

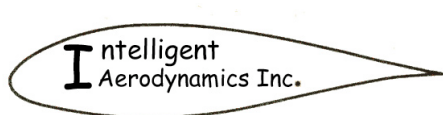


Supplementary Data



Appendix 1

Further Data for Part 2



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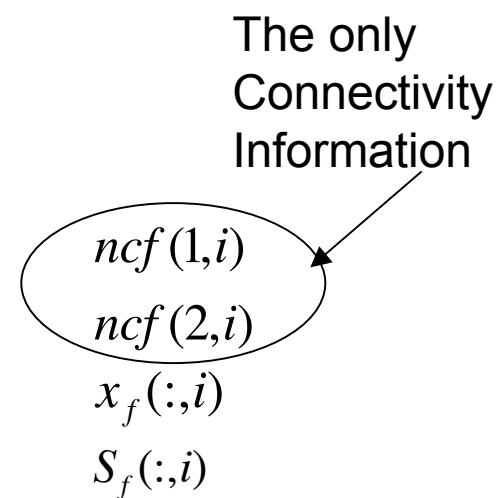
Copyright 2004, A. Jameson, G. Singh, G. May, and K. Leoviriyakit



What do We Need For a Face-Based Data Structure?

List of faces

- Pointer to the two cells separated by the face
- Coordinates of face centroid
- Directed face area



List of cells

- Coordinates of cell centroid $x_c(:,i)$
- Volume $Vol(i)$
- Averaged values of solution variables $W(:,i)$

List of boundary faces and boundary conditions applied to each face



The Main Algorithm

```
do i=ncell1,ncell2
```

```
  set residual(i) to zero
```

```
end do
```

```
do n=ntface1,ntface2
```

```
  A = ncf(1,n)
```

```
  B = ncf(2,n)
```

```
  flux(N) = f(solution(A),solution(B))
```

```
  residual(B) = residual(B) + flux(N)
```

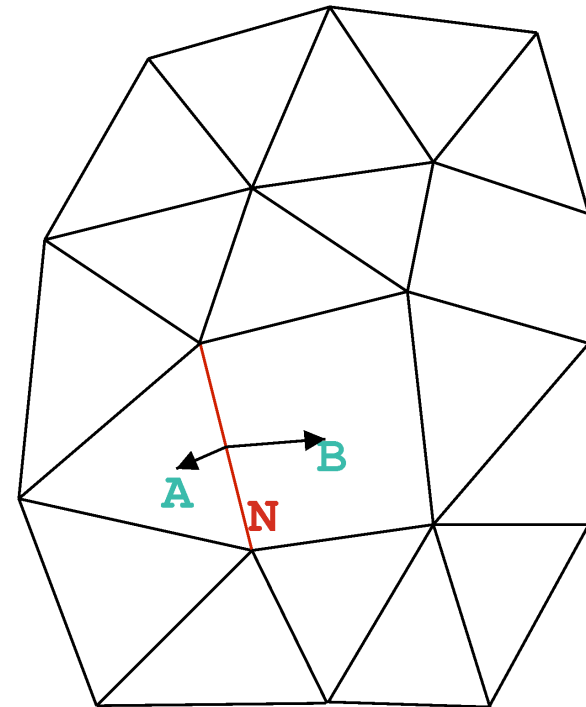
```
  residual(A) = residual(A) - flux(N)
```

```
end do
```

```
do i=ncell1,ncell2
```

```
  solution(i) = solution(i) - residual(i)
```

```
end do
```



Constructing the Fluxes

$$f(\vec{x}, \vec{u}, t, \Delta) = \int_0^t g(\vec{x} - \vec{u}(t - \tau), \vec{u}, t, \Delta) e^{-\frac{(t-\tau)}{\Delta}} d\tau + e^{-\frac{t}{\Delta}} f_0(\vec{x} - \vec{u}t, \vec{u}, \Delta)$$

- The initial nonequilibrium distribution at a face can be approximated from an approximate distribution in the neighboring cells A and B

$$f_0 = \begin{cases} g^A(1 + a^A s), & s < 0 \\ g^B(1 + a^B s), & s \geq 0 \end{cases} \quad s = (\vec{x} - \vec{x}_f) \cdot \vec{n}_f$$

- Similarly, the time dependent equilibrium distribution g at the cell interface is obtained as

$$g = g_0 \left\{ 1 + (1 - H(s)) \tilde{a}^A s + H(s) \tilde{a}^B s + \bar{A}t \right\}, \quad H(x) = \begin{cases} 0, & x < 0 \\ 1, & x \geq 0 \end{cases}$$



Constructing the Fluxes

- The Fluxes, in x say, are computed as

$$F = \begin{bmatrix} F_x \\ F_u \\ F_v \\ F_w \\ F_E \end{bmatrix} = \begin{bmatrix} \rho \\ \rho u \\ \rho v \\ \rho w \\ \frac{1}{2} \rho (u^2 + v^2 + w^2 + p) \end{bmatrix} \begin{bmatrix} 1 \\ u \\ v \\ w \\ (u^2 + v^2 + w^2 + p) \end{bmatrix} f(x_f, t, u, v, w, p) dudvdw$$

- Compute the fluxes in y and z (G,H) in similar fashion
- Update the solution according to

$$\vec{W}^{n+1} = \vec{W}^n + \frac{1}{V} \int_0^t \vec{F} \cdot d\vec{A} dt \quad \vec{F} = (F, G, H)^T$$



Appendix 2

Shape Optimization Results via Control Theory



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Redesign of the Boeing 747 Wing at its Cruise Mach Number

- Constraints :
- Fixed $C_L = 0.42$
 - Fixed span-load distribution
 - Fixed thickness

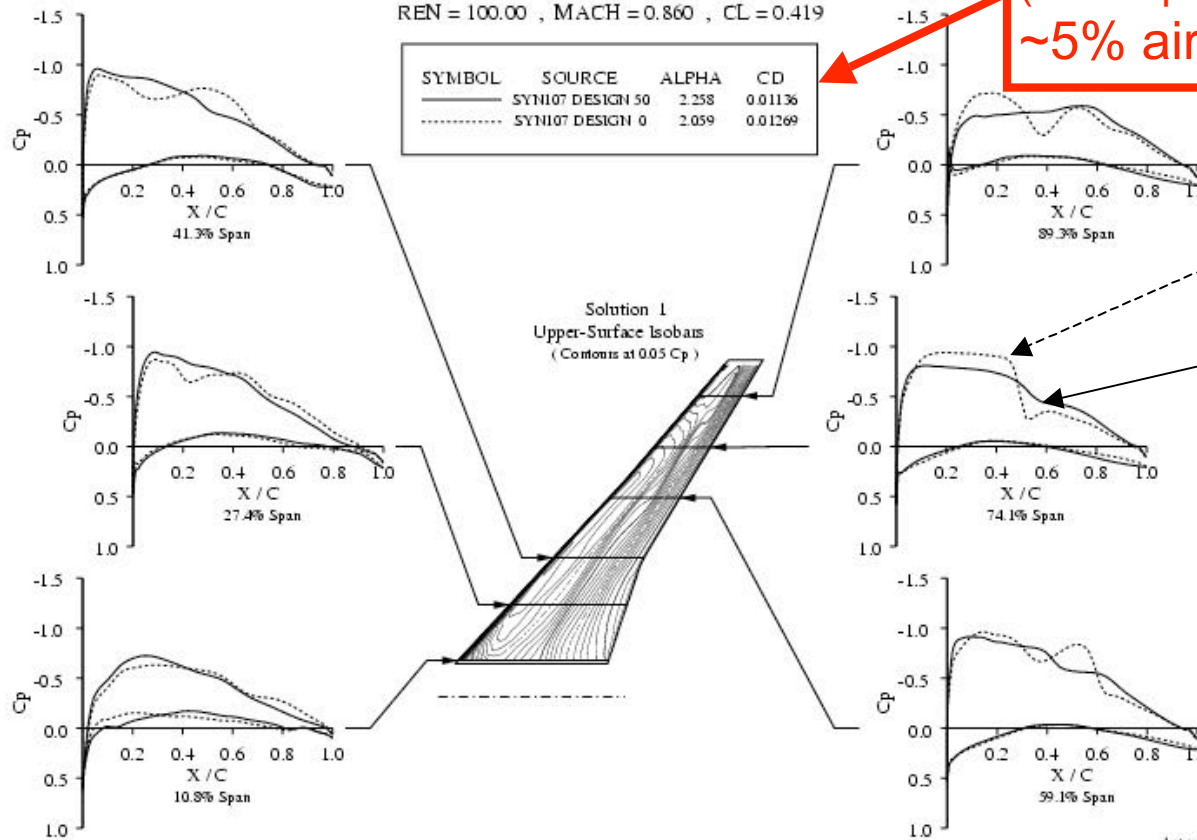
RANS Calculations

COMPARISON OF CHORDWISE PRESSURE DISTRIBUTIONS
B747 WING-BODY

REN = 100.00 , MACH = 0.860 , CL = 0.419

SYMBOL	SOURCE	ALPHA	CD
—	SYN107 DESIGN 50	2.258	0.01136
---	SYN107 DESIGN 0	2.059	0.01269

10% wing drag saving
(3 hrs cpu time - 16proc.)
~5% aircraft drag saving



baseline
redesign

COMPPLOT

Antony Jameson
14:40 Tue
28 May 02

Intelligent
Aerodynamics Inc.

Aerospace Computing Laboratory

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132

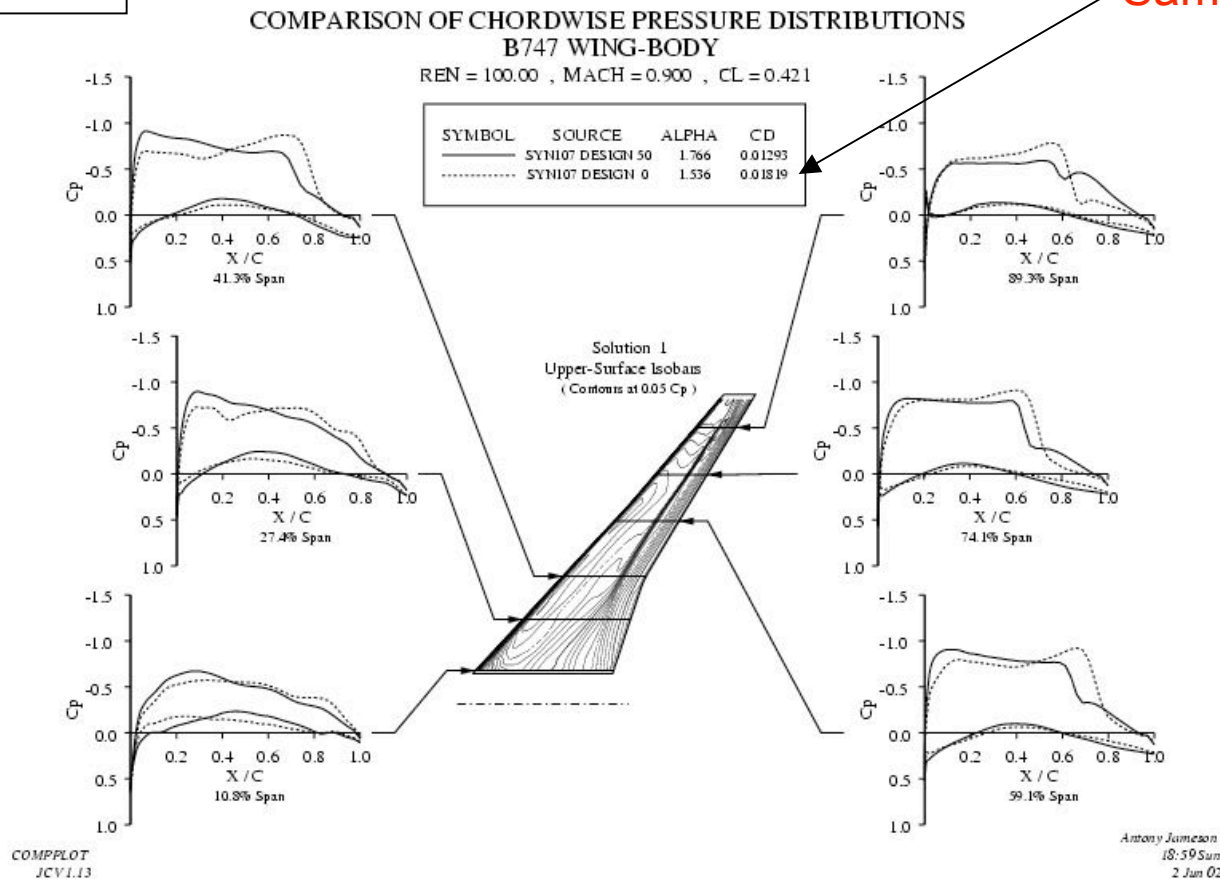


Redesign of the Boeing 747 Wing at Mach 0.9 “Sonic Cruiser”

- Constraints :
- Fixed $C_L = 0.42$
 - Fixed span-load distribution
 - Fixed thickness

RANS Calculations

Same C_D @Cruise



We can fly faster at the same drag.



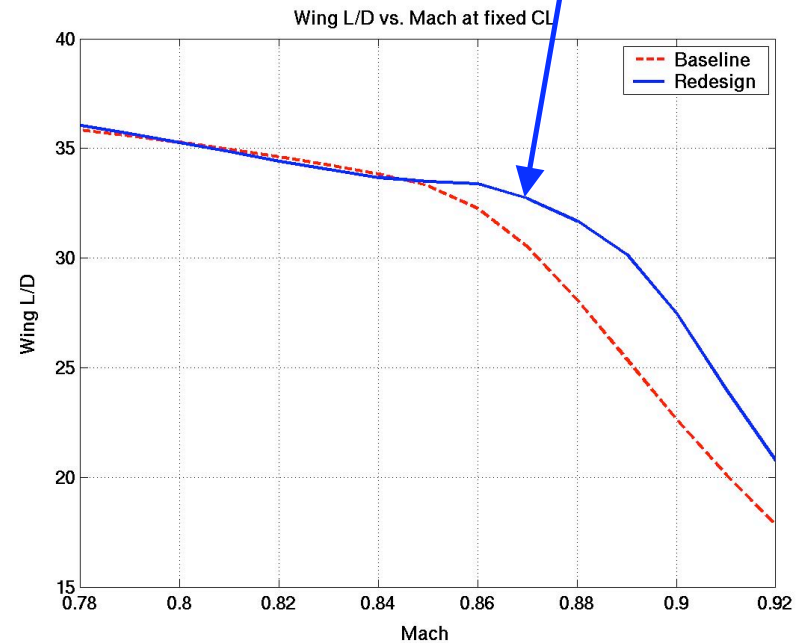
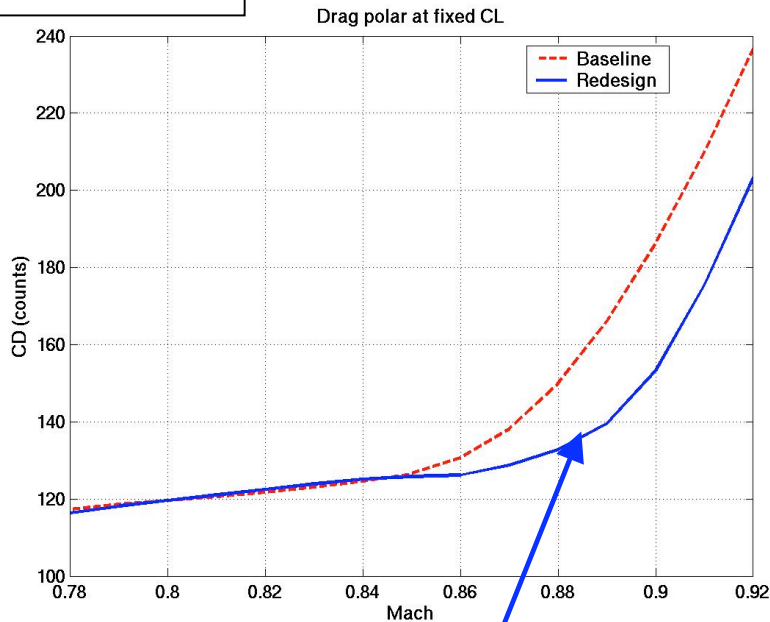
Redesign of the Boeing 747: Drag Rise

(Three-Point Design)

- Constraints :
- Fixed $C_L = 0.42$
 - Fixed span-load distribution
 - Fixed thickness

Improved Wing L/D

RANS Calculations



Improved M_{DD}

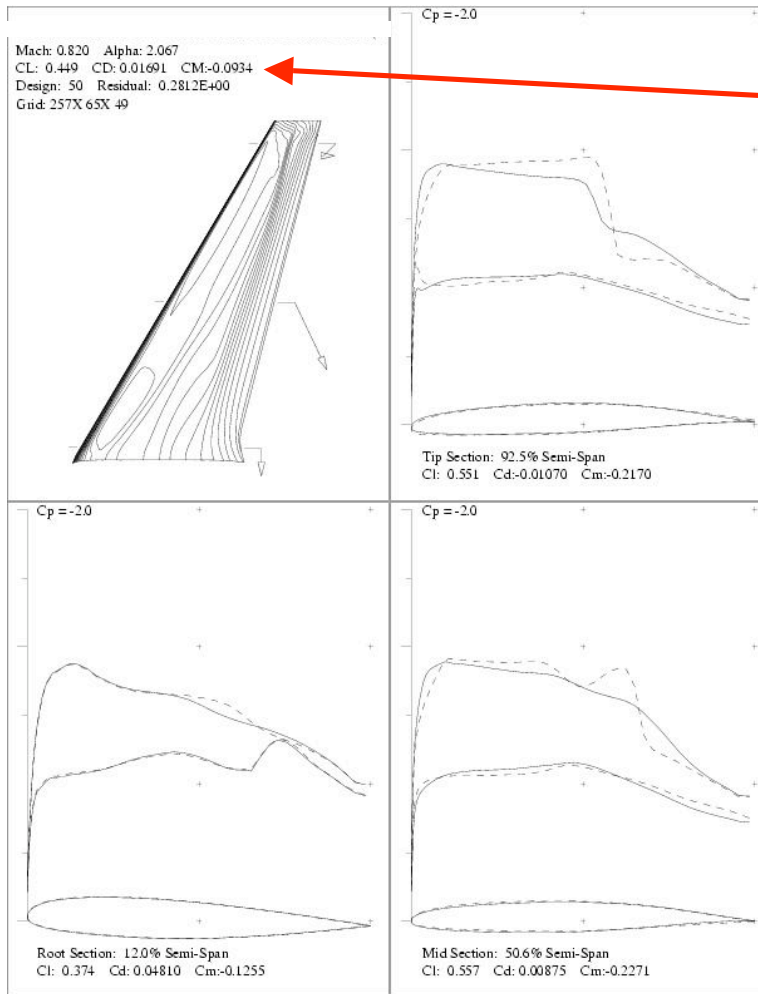
- benefit → Lower drag at the same Mach Number
- benefit → Fly faster with the same drag



Redesign of an Executive Jet

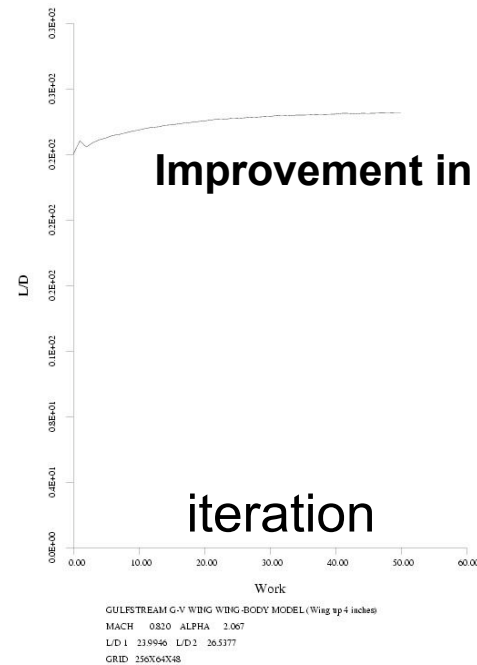
- Constraints :
- Fixed $C_L = 0.45$
 - Fixed span-load distribution
 - Fixed thickness

RANS Calculations



Drag saving:
189 counts to 169 counts

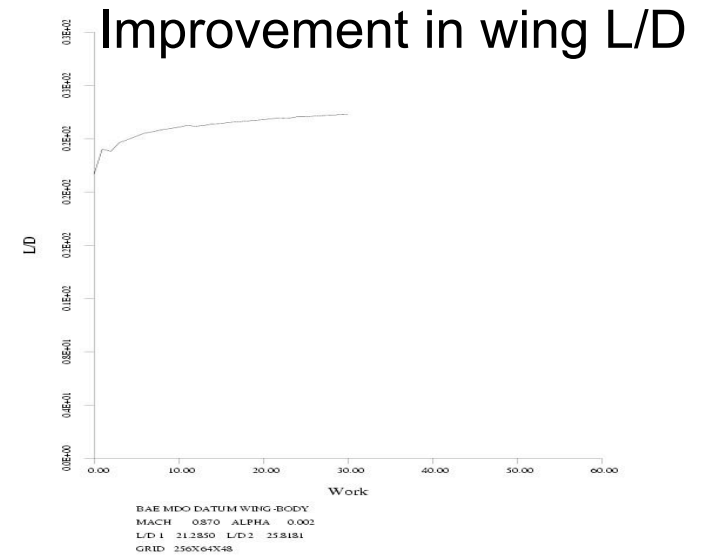
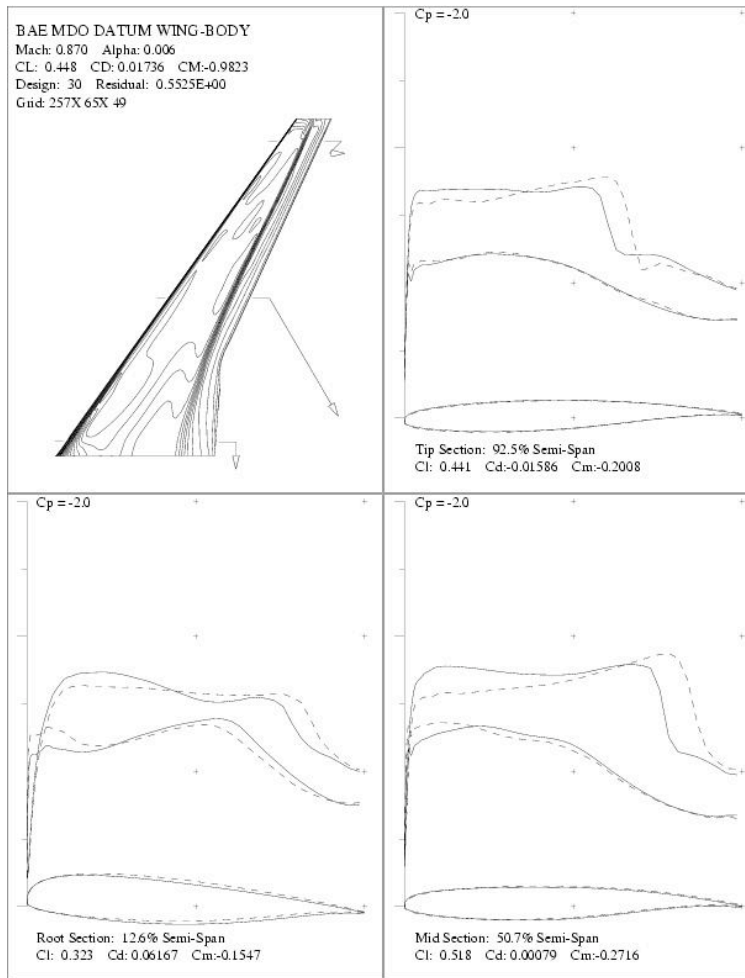
Wing L/D



Redesign of the BAE MDO Datum Wing-Body

- Constraints :
- Fixed $C_L = .454$
 - Fixed span-load distribution
 - Fixed thickness

RANS Calculations

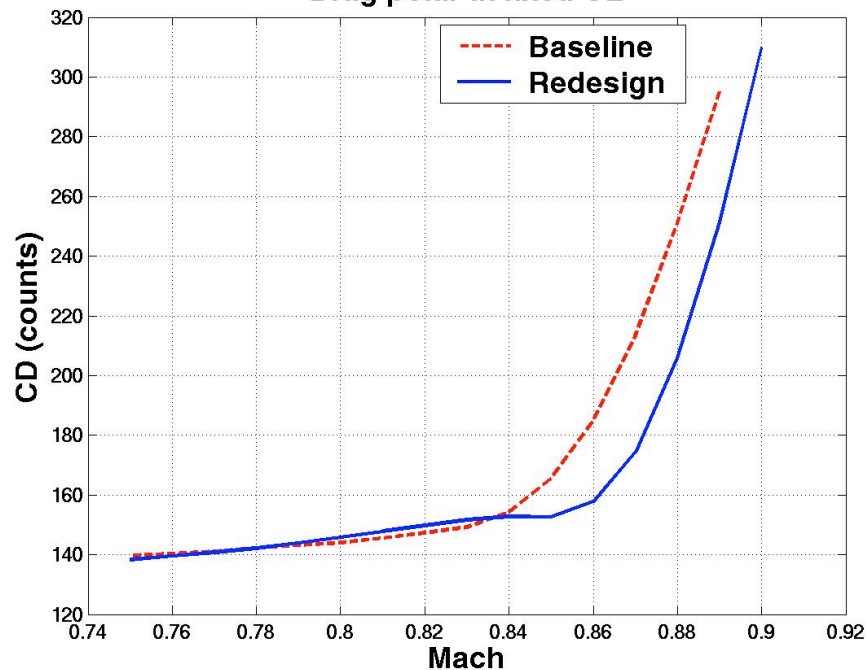


Redesign of the BAE MDO Datum Wing-Body : Drag Rise

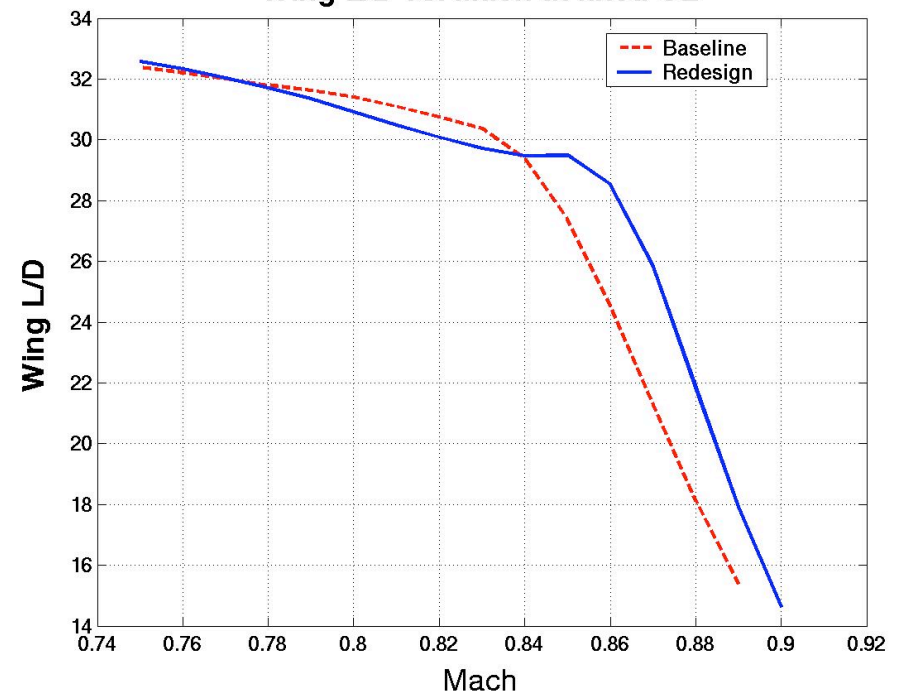
- Constraints :
- Fixed $C_L = 0.454$
 - Fixed span-load distribution
 - Fixed thickness

RANS Calculations

Drag polar at fixed C_L



Wing L/D vs. Mach at fixed C_L



ML/D improved by 3 %



Theses on BGK Method

- *Chongam Kim*
- *Yee Feng Ruan*
- *Balaji Srinivasan*
- *Georg May*



Theses on Dual Time-Stepping and Time-Spectral Method

- *Juan Alonso*
- *Andre Belov*
- *Bing Ham Liou*
- *Paul Lin*
- *Sriram Shankaran*
- *Mathew McMullen*
- *Arathi Gopinath*



Theses on Shape Optimization

- *James Reuther*
- *Sangho Kim*
- *Siva Nadarajah*
- *Sriram Shankaran*
- *Kasidit Leoviriyakit*

(At Princeton under Luigi Martinelli)

- *James Dreyer*



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Analysis and Design of Numerical Schemes for Gas Dynamics 1 Artificial Diffusion, Upwind Biasing, Limiters and Their Effect on Accuracy and Multigrid Convergence, RIACS Technical Report 94.15, International Journal of Computational Fluid Dynamics, Vol. 4, 1995, pp. 171-218.

Analysis and Design of Numerical Schemes for Gas Dynamics 2 Artificial Diffusion and Discrete Shock Structure, RIACS Report No. 94.16, International Journal of Computational Fluid Dynamics, Vol. 5, 1995, pp. 1-38.

Parallel Computations of Unsteady Incompressible Viscous Flows with a Fully-Implicit Multigrid Driven Algorithm (with A. Belov, L. Martinelli) Proceedings of 6th International Symposium on Computational Fluid Dynamics, Lake Tahoe, September 1995.

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Time-Accurate Simulation of Helicopter Rotor Flows Including Aeroelastic Effects (with S. Sheffer, J. Alonso, L. Martinelli), AIAA 97-0399, AIAA 35th Aerospace Sciences Meeting and Exhibit, Reno, January 1997.

An Accurate LED-BGK Solver on Unstructured Adaptive Meshes (with C.A. Kim, K. Xu, L. Martinelli), AIAA 97-0328, AIAA 35th Aerospace Sciences Meeting and Exhibit, Reno, January 1997.



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Analysis and Implementation of the Gas-Kinetic BGK Scheme for Computational Gas Dynamics (with K. Xu, C. Kim, L. Martinelli) International Journal of Numerical Methods in Fluids, Vol. 25, No. 1, July 1997, pp. 21-49

A Robust and Accurate LED-BGK Solver on Unstructured Adaptive Meshes, C. Kim, A. Jameson, Journal of Computational Physics, Vol. 143, March 1998, pp. 598 - 627.

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1982 - 1997	<i>James S. McDonnell Distinguished University Professor of Aerospace Engineering</i> , Princeton University
1980 - 1982	<i>Director, Program in Applied and Computational Mathematics</i> , Princeton University

Honors

2002	Docteur Honoris Causa, Uppsala University
2001	Docteur Honoris Causa , Université Pierre et Marie Curie, Paris VI
1997	Foreign Associate, National Academy of Engineering
1995	ASME Spirit of St. Louis Medal
1995	Fellow of the Royal Society of London
1993	American Institute of Aeronautics and Astronautics Fluid Dynamics Award
1990	Fellow of the American Institute of Aeronautics and Astronautics
1988	Gold Medal of the Royal Aeronautical Society
1980	NASA Medal for Exeptional Scientific Achievement

Bio Antony Jameson

Biography

1953 - 1955	2nd Lieutenant British Army in Malaysia
1955	Bristol Siddeley Jet Engines
1955 - 1963	Cambridge university BA, MA Mechanical Sciences PhD Magneto-Hydrodynamics
1964 - 1965	Economist London
1965 - 1966	Chief Mathematician
1966 - 1972	Grumman Aerospace Aerodynamics Department
1972 - 1980	Courant Institute of Mathematics Professor of Computer Science 1974-80
1980 - 1997	Princeton University James S. McDonnell Distinguished University Professor of Aerospace Engineering 1982-97
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Relationship to the Boeing Company

- Antony Jameson has had a long standing relationship with both McDonnell Douglas and Boeing
- Flo22, provided to Douglas in 1976, was a principal design tool for the C17 wing design. It is still used by Phantom Works.
- Flo27 was incorporated as the flow solver in the Boeing A488 software, which was used in the wing design of the Boeing 757, 767, and 777.
- Flo27 was incorporated in the McDonnell Douglas DACTRAN 10, 20 and 30 codes
- The AIRPLANE code was used in the MD11 CPIP, HSR and C17 winglet studies

