

A Methodology for Topology Optimization in Fluid Dynamics

Fluid Dynamics

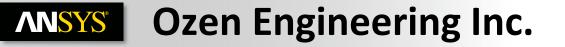
Structural Mechanics

Electromagnetics

Systems and Multiphysics

Chris Cowan

Ozen Engineering, Inc. 1210 E. Arques Ave, Suite 207 Sunnyvale, CA 94085 info@ozeninc.com



We are your local ANSYS channel partner!



With over 25 years of experience in Finite Element Simulations and Engineering Consulting, we collaborate with customers to provide the best in class expertise and solutions to their problems, enabling them to succeed.

www.ozeninc.com

ANSYS Contents & Abstract

- 1. Introduction
- 2. Description of Problem
- 3. Geometry Creation
- 4. Modeling
- 5. Optimization

A CFD case study is presented to demonstrate a methodology of topology optimization. A flow cell with non-uniform thickness profile is prepared in ANSYS DesignModeler using parametric surfacing techniques. Flow characteristics and uniformity are evaluated using ANSYS CFX. Geometric design variations are explored then refined to satisfy performance criteria using Design of Experiments and optimization routines in ANSYS DesignXplorer.



Demonstrate a methodology to optimize fluid behavior by introducing a variable thickness shape profile to a flow channel.

The focus of this discussion is on:

- Preparing a model to solve quickly and accurately without failure for a large set of parameters.
- Detailing a typical optimization procedure

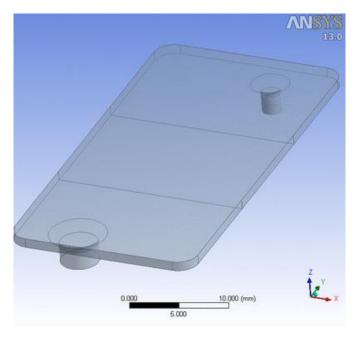
This methodology is relevant for many industrial applications and can be adapted to different types of simulation. It has been demonstrated to significantly enhance performance for biomedical equipment and fuel cells.

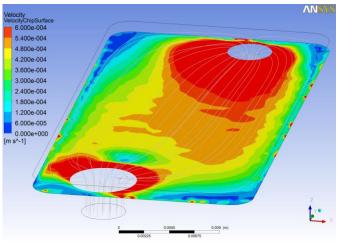
ANSYS Description of Problem

Fluid passes through a flow cell. Properties of the fluid are measured on the sensor surface. To obtain accurate measurements, the fluid should have a uniform velocity.

The baseline design features a constant thickness channel with circular inlet & outlet.

Optimize the geometry by applying a nonuniform thickness profile along the top surface of the flow cell. The objective is to improve sensor performance by increasing flow uniformity.

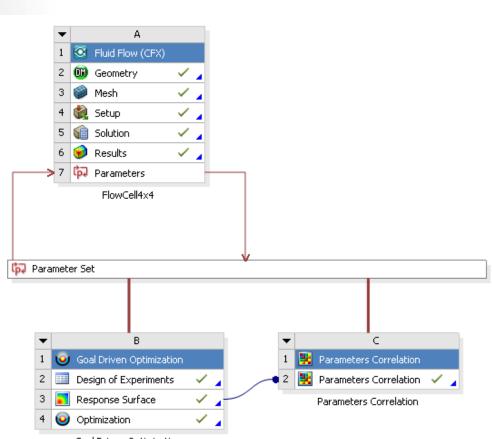




ANSYS Workbench Solution Approach

Parametric geometry is created using DesignModeler.

- A mesh with inflation boundary is created using Workbench Meshing
- The fluid model is setup, solved, and post-processed using CFX.
- The design space is explored using DesignXplorer and parameters are refined to improve performance.

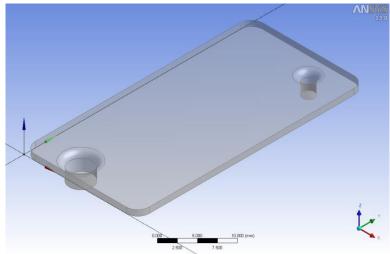


Goal Driven Optimization

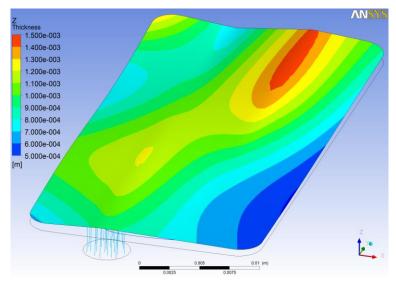
ANSYS Geometry - Overview

Parametric geometry creation relies on a few simple operations:

- Offset Sketching Planes
- Skin/Loft
- Merge Surface
- Boolean Unite
- Blend
- Named Selections

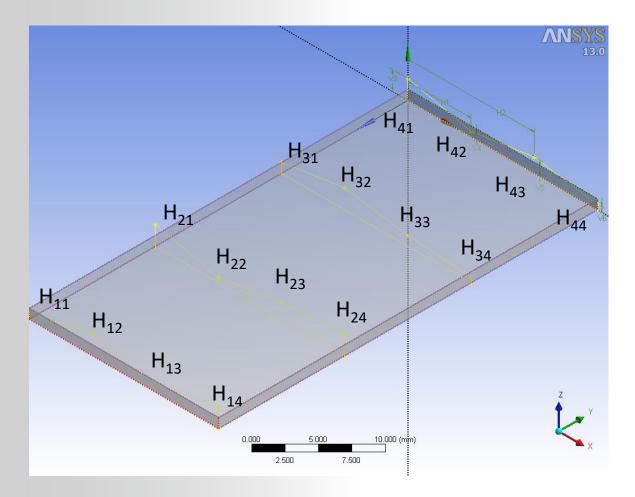


Baseline template model



Contour plot showing channel thickness

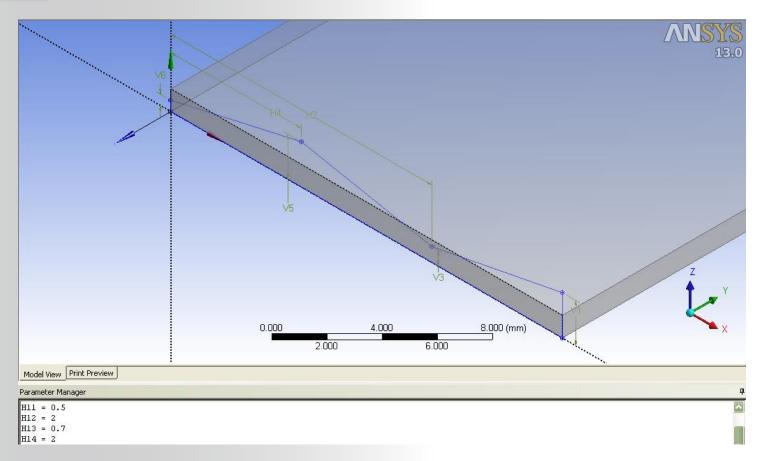
ANSYS Geometry - Approach



The variable thickness fluid domain is created using a 4x4 array of evenly-spaced control points which are sketched over the template geometry.

16 parameter names are assigned to thickness control points (named Hij where 1≤i≤4 and 1≤j≤4).

ANSYS Geometry - Sketching Planes



Face-boundary outline plane features fixed lines (black-dashed) from the scoped geometry.

Non-parametric features are spaced constrained to existing solid edge outlines.

Thickness control points are dimensioned from the sensor surface (bottom) and parameterized.

ANSYS Geometry - Skin/Loft

| 🔯 A: FlowCell4x4 - DesignModeler | | - 6 |
|---|--|--|
| File Create Concept Tools View Help | | |
|) 🖉 🔜 🛃 📫 🕽 Olndo @Redo 🛛 Select: 🎠 🍢 🕅 🕅 💽 | | |
| II · h · h · h · h · h | | |
| Array4 • 🖈 Profile5 • 🙋 | | |
| | /Loft 🖪 Thin/Surface 💊 Blend 🗸 🦠 Chamfer 🔗 Point 📴 Parameters | |
| A CONTRACTOR OF | Graphics | 4 |
| A: FlowCell4x4 | | ANSYS |
| XYPlane | and the second | |
| | and the second | 13.0 |
| Template | and the second sec | |
| | | |
| 日本 Array1 | | The second secon |
| Profile1 قصر | | HO |
| - Array2 | | |
| ل المربية Profile2 ⊡ | | |
| → Arrays | | |
| | | V4 The second second |
| Profile5 | | |
| E⊢ & Skin1 | | |
| ୍କୁ ଅ Profile1 କୁ ଅ Profile2 | | V5 |
| - 20 Profile4 20 Profile5 | | |
| SmoothSurface | | |
| BlendCorners | | VETTIN |
| - X Inlet | | |
| | | |
| BlendInletOutlet | | |
| x SensorSurface | | |
| 2 Parts, 4 Bodies | | |
| Sketching Modeling | | |
| Details View 7 | | |
| Details of Profile5 | | |
| Sketch Profile5 | | |
| Sketch Visibility Show Sketch | | |
| Show Constraints? No | | |
| Dimensions: 6 D H1 6.6667 mm | | |
| D H2 13.333 mm | | |
| D V3 2 mm | | 7 |
| D V4 0.5 mm | | |
| D V5 1.5 mm | | Т Y |
| D V6 0.5 mm | | <u> </u> |
| Edges: 6 | | <pre></pre> |

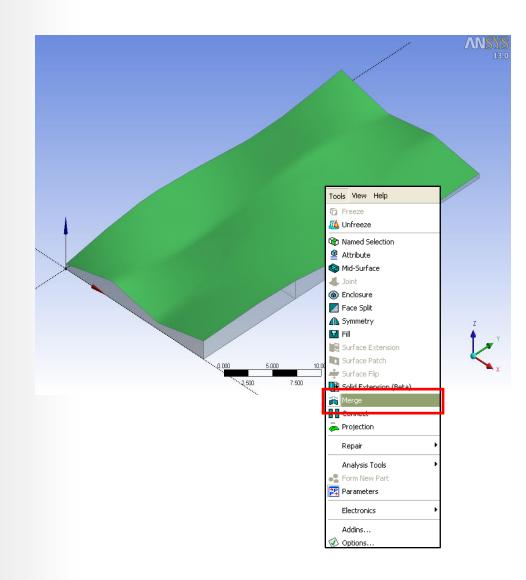
Loft through a series of profiles on different planes.

This enables a robust part with parameterized thicknesses.

ANSYS Geometry – Surface Merge

The multi-faceted surface is smoothed by merging faces.

- Typically, the merge operation is used to remove sharp edges and simplify geometry for meshing (virtual topology).
- Alternate methods lead to a high rate of DOE failure.
- Penetration at thin crosssections.
- Spline orientation





Boolean Unite:

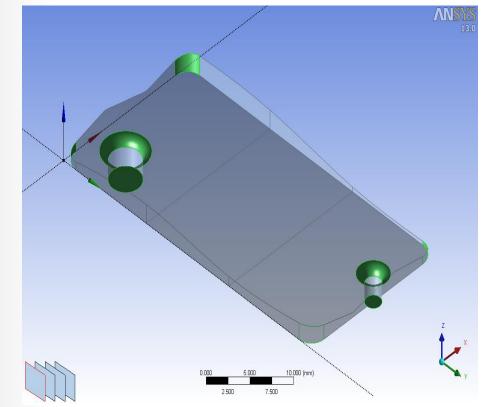
- Merges the inlet/outlet tubes with the channel
- Single part, continuous mesh, no domain interfaces required

Blend Edges:

- Creates smooth transitions.
- Should follow the skin/loft operation.

Named Selections:

- Simplify setup operations
- Automatic mesh inflation
- CFX-Pre boundary conditions
- CFX-Post results evaluation
- Specified Inlet, Outlet, and SensorSurface

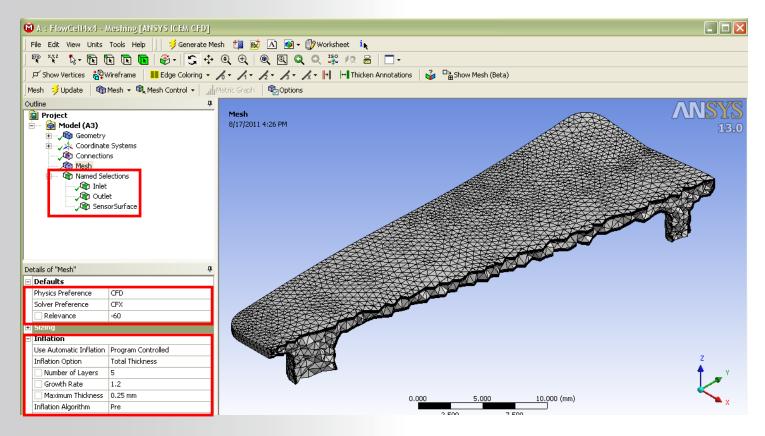


ANSYS Meshing

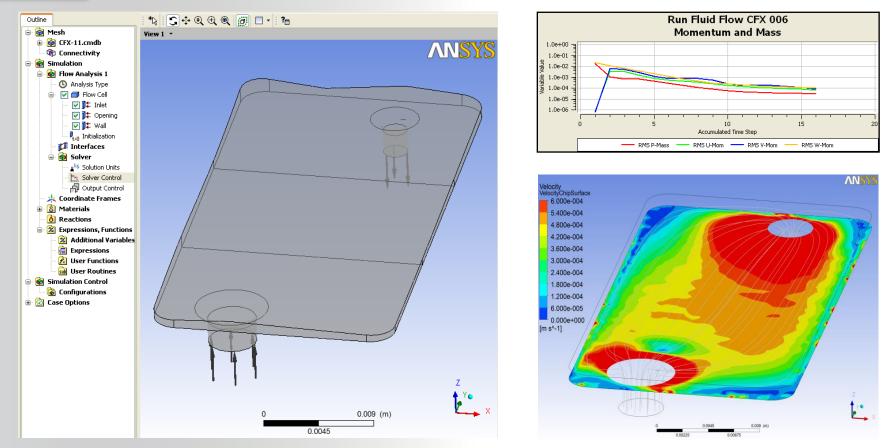
Default settings for the CFD physics preference, with automatic inflation enabled.

Named selection settings are modified for program controlled inflation. "SensorSurface" is included and "Inlet/Outlet" are excluded as boundaries.

Balance conflicting objectives of obtaining reasonable solution accuracy vs. very short solution times by running a mesh convergence study.



ANSYS CFD – Setup, Solution, Results



A simple steady-state fluid analysis is setup in CFX-Pre.

The analysis solves in less than 1 minute in CFX-Solver on a mid-range workstation.

The objective of this optimization is to obtain a uniform fluid velocity along the sensor surface. The metric used to quantify uniformity is standard deviation of fluid velocity.

ANSYS CFD – Standard Deviation

- Standard deviation provides a measurement of result variation from the mean. Uniform flow corresponds to minimized standard deviation of velocity.
- Standard deviation is not a standard function within CFD-Post.
- **Evaluate using expressions:**
- 1. Calculate the number of nodes in the region.

```
StDev Count = count()@SensorSurface
```

2. Calculate the mean value of the desired variable.

StDev Mean = sum(Velocity)@SensorSurface / StDev Count

3. Define variance as the squared difference between the variable and the mean value.

StDev Variance = sum((Velocity - StDev Mean)^2)@SensorSurface / (StDev Count - 1)

4. Define standard deviation as the square root of variance.

StDev Sensor Velocity = sqrt(StDev Variance)

Additional information on these expressions is available on the ANSYS Customer Portal.

ANSYS Optimization - Summary

Optimization is implemented using ANSYS DesignXplorer, which supports any analysis type or CAD available to Workbench.

Optimization falls under the broad approach of design exploration – the process of understanding the relationship between design inputs and response outputs. Common tasks include:

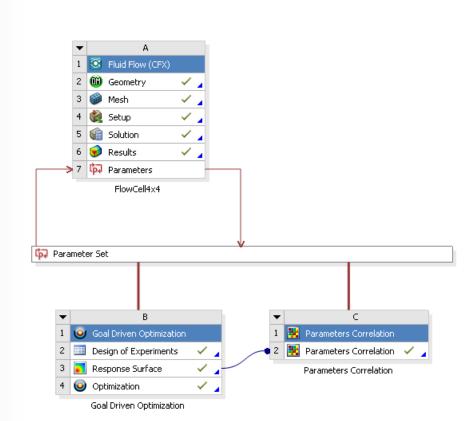
- What-if studies
- Design of experiments
- Response surface modeling
- Min-max search

- Parameter correlation
- Goal driven optimization
- Six Sigma

Optimization expands on results obtained through design exploration by predicting and verifying designs to satisfy goals using response surface methodology.

ANSYS Optimization – Typical Workflow

- **1. Generate and solve DOE**
- 2. Generate response surface(s) and review Goodness of Fit
- 3. Review Parameters Correlation
- 4. Screen to select preliminary candidates
- 5. Optimize to refine candidates
- 6. Solve design points to validate leading candidates



ANSYS Design of Experiments

Design of Experiments (DOE) refers to the structured generation of a set of data used to gain understanding on the relationship between design variables.

A minimum number of DOE solutions are required by DesignXplorer to build a response surface: 2 variables → 9 samples 8 variables → 81 samples 16 variables → 289 samples

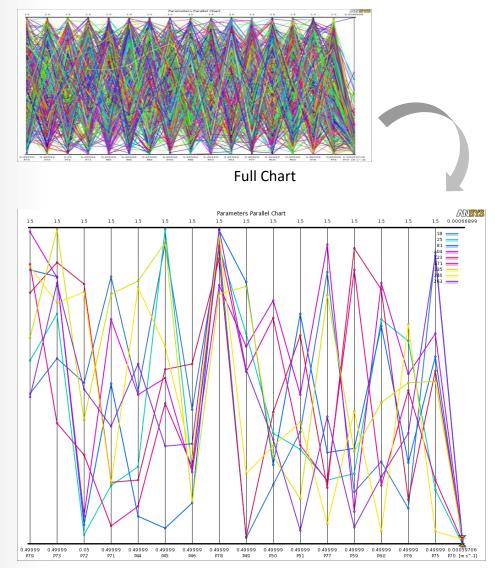
ANSYS Design of Experiments

Specify ranges for all input parameters: 0.3mm ≤ Thickness ≤ 1.5mm

Select the DOE methodology: Optimal Space-Filling Central Composite Design Custom

Samples are distributed throughout the design space.

Review DOE results using Parameters Parallel Charts

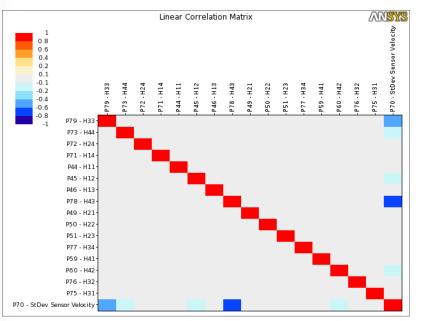


Filtered to best results

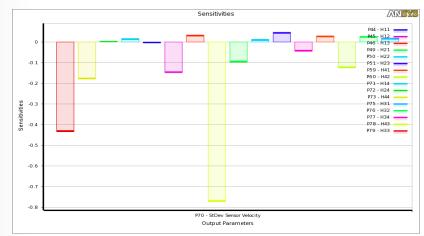
ANSYS Parameter Correlation

Visualize importance and nature of variable relationships

- Mathematical sampling methods are used to identify parameter relationships for inputs and outputs.
- Remove less-important input parameters from design exploration
- Reduce the size of the DOE
- Increase speed and accuracy of surface response generation
- Use existing DOE results or autogenerate new samples



Correlation matrix



Output sensitivities

ANSYS Response Surface

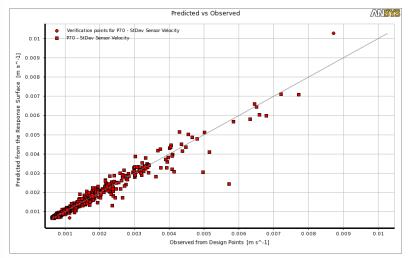
The meta model approximates the relationship between inputs and outputs by curve-fitting DOE sampling data.

Benefit: Rapidly predict results for theoretical designs without solving hard points.

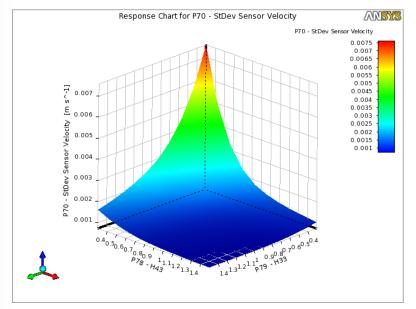
Response surface training methodology:

Second-Order Polynomial

Kringing, Non-Parametric Regression, Neural Network Review & refine Goodness-of-Fit



Second-order polynomial (Goodness-of-Fit 0.95)



ANSYS Goal Driven Optimization

A technique to obtain the best designs from a sample set by evaluating theoretical inputs using response surface methodology. Weighted parameter guidance sets optimization goals and rules.

Specify objectives for output and input parameters:

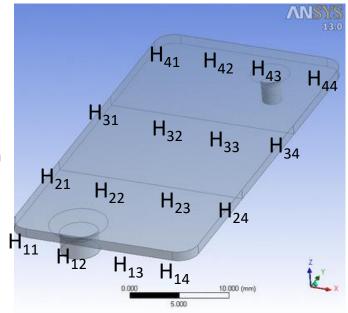
- Minimize, Maximize, Relative to target
- Prioritize objectives
- Specify initial values

Screening:

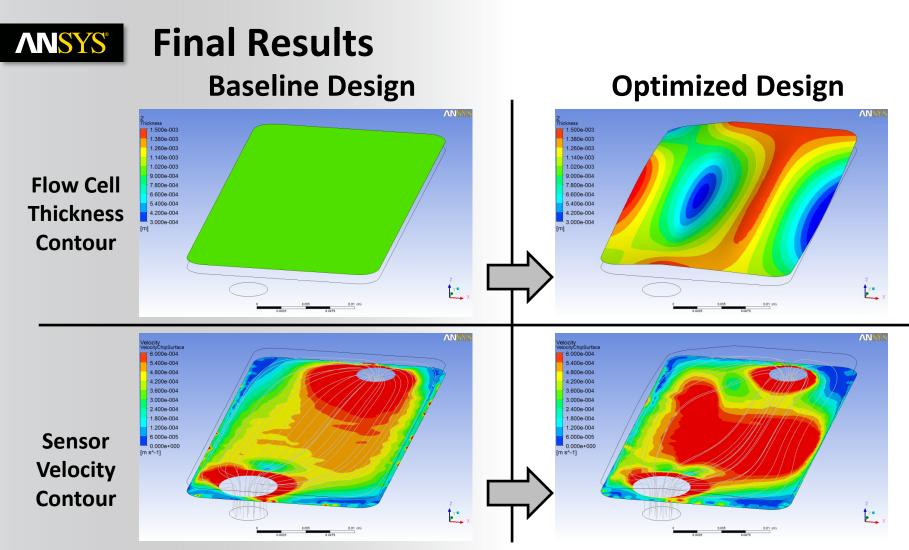
- Overview of design space using random number generation
- Multiple design optimization candidates

NLPQL:

- Single-objective gradient-based optimizer
- Prone to local minima
- Single optimization design candidate



| Table of Schematic I4: Optimization | | | | | | | | | | | | | | | | | | |
|-------------------------------------|----------------------|-------------|-----------|-----------|-----------|-----------|-----------|-----------|-------------|-----------|-----------|-----------|----------------|-----------|-----------|-----------|-----------|--------------------------------------|
| | A | В | С | D | E | F | G | н | I | J | к | L | м | N | 0 | Р | Q | R |
| 1 | | P79 - H33 | P73 - H44 | P72 - H24 | P71 - H14 | P44 - H11 | P45 - H12 | P46 - H13 | P78 - H43 | P49 - H21 | P50 - H22 | P51 - H23 | P77 - H34 | P59 - H41 | P60 - H42 | P76 - H32 | P75 - H31 | P70 - StDev Sensor Velocity (m s^-1) |
| 2 | | | | | | | | | | | | | | | | | | |
| 3 | Initial Value | 1.498 | 1.4982 | 0.3 | 0.3 | 1.1 | 1.4982 | 0.5 | 1.4948 | 1.45 | 0.5 | 0.5 | 1.3 | 0.4 | 1.498 | 0.4 | 0.4 | |
| 4 | Objective | NO OD 💽 | NO O 💽 | NO O 💽 | NO O 💽 | NO O 💽 | NO O 💽 | NO O 💌 | NO CO 💽 | NO O | NO OD 💌 | NO U 💽 | No Objective 💽 | NO U 💽 | NO O 💌 | NO U 💽 | NO O 💽 | Minimize |
| 5 | Target Value | | | | | | | | | | | | | | | | | |
| 6 | Importance | Default 🗾 💌 | Default 💽 | Default 💽 | Default 💽 | Default 🖃 | Default 💽 | Default 💽 | Default 🔄 💽 | Default 💽 | Default 💽 | Default 💽 | Default 💽 | Default 💽 | Default 💌 | Default 💽 | Default 💽 | Higher 🗾 |
| 7 | 7 = Candidate Points | | | | | | | | | | | | | | | | | |
| 8 | Candidate A | - 1.4983 | - 1.4 | - 0.2 | - 0.3 | - 1.1 | - 1.4 | - 0.5 | - 1.4948 | - 1.1 | - 1.4983 | - 0.3 | - 0.3 | - 0.3 | - 1.4 | - 0.3 | - 1.4983 | A 0.00019672 |
| 9 | Verification A | - 1.4903 | 1.4 | 0.2 | 0.5 | 1.1 | 1.4 | 0.5 | 1.4940 | 1.1 | 1.4903 | 0.5 | 0.5 | 0.3 | 1.4 | 0.3 | 1,4903 | A 0.00057896 |



The contoured thickness profile of the optimized design leads to 40% improvement in flow uniformity at the sensor surface (StDev 0.57 mm/s compared to 0.91 mm/s).

Additional refinement iterations and more extensive parameter ranges can be incorporated to improve results further.



This presentation has demonstrated how ANSYS products & technology can be used to create robust parametric models and solve shape optimization for a fluid domain.

These results show that parametric modeling and optimization techniques can be employed to rapidly and accurately refine a product design to include amorphous features which improve flow characteristics.



QUESTIONS ? Thank you for your attention

Fluid Dynamics

Structural Mechanics

Electromagnetics

Systems and Multiphysics

Ozen Engineering Inc: 1210 E. Arques Ave #207 • Sunnyvale, CA 94085 • (408) 732-4665

