

# Over-Wing Nacelle (OWN) Placement Evaluation

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Cost Share Partner: Georgia Institute of Technology

## Objective:

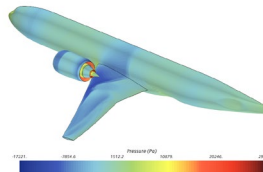
- Use multidisciplinary design analysis and optimization (MDO) methods to assess environmental impact of over-wing nacelle (OWN) placement
- Emphasis on high fidelity aerodynamics to capture drag penalty

## Project Benefits:

- Enable accurate tradeoffs between noise benefits and fuel burn penalties
- Demonstrate computationally efficient methods for aircraft design studies

## Research Approach:

- Computational efficiency is key challenge
- Numerical uncertainty is a major theme
- Fair comparison requires two optimization processes: OWN vs under-wing nacelle (UWN)



## Major Accomplishments (to date):

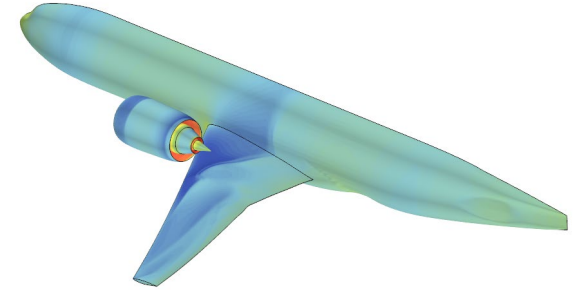
- Developed coupled aero-propulsion analysis method
- First stage nacelle location study for OWN vs UWN

## Future Work / Schedule:

- Larger scale, second-stage optimization on larger set of shape variables
- Final deliverable in June 2022

# Do OWN Benefits Outweigh Disadvantages?

- Focus is **forward-mounted OWN** for single-aisle transports

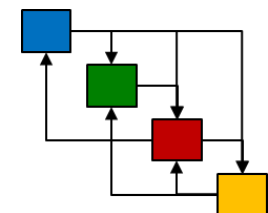
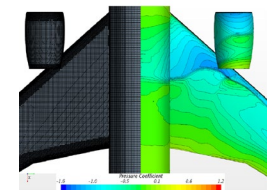


- Clearance for larger, efficient engines
- Noise shielding
- *Complex aero-propulsion coupling and interference*



- **Technical challenges**

- CFD computational cost
- Multi-disciplinary design analysis and optimization (MDAO)

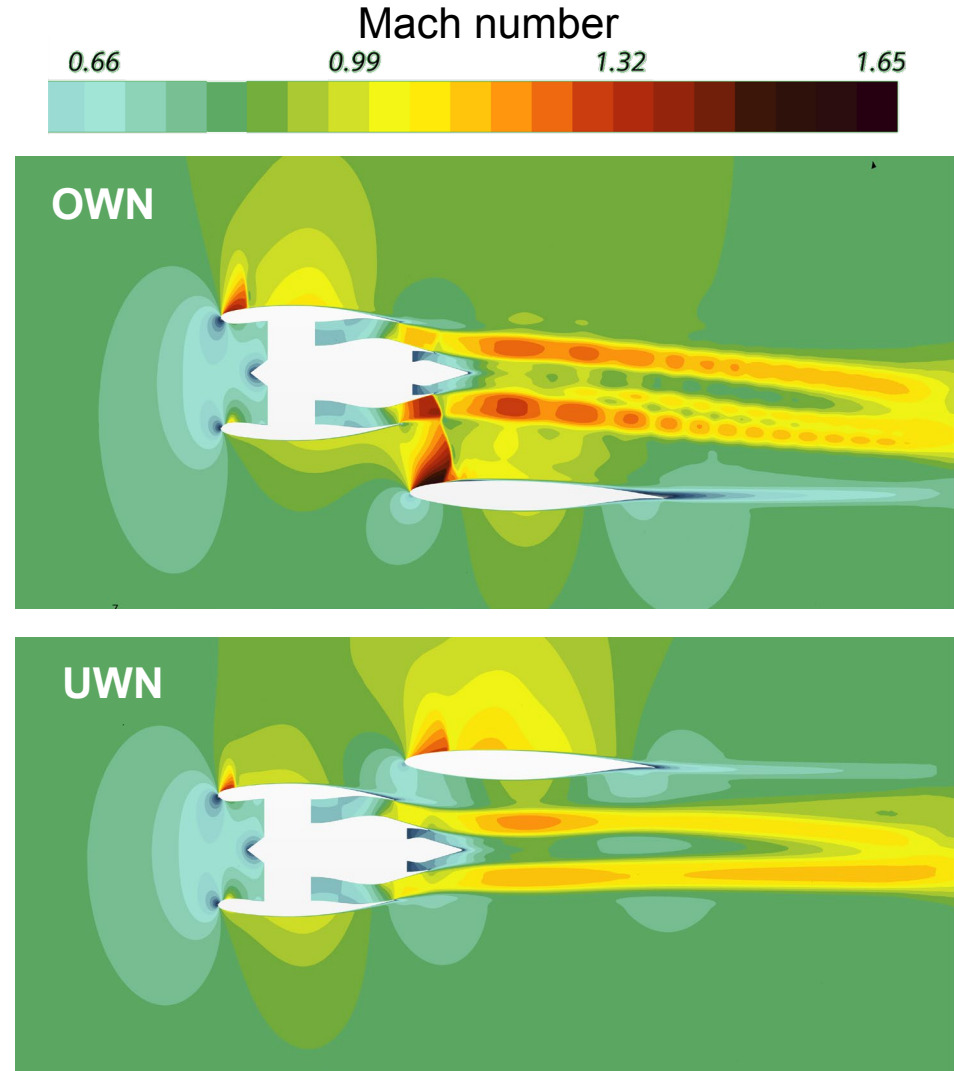


# Methodology: Optimize OWN and UWN Under Same Rules

- More physics constraints already enforced in mature UWN
- Challenge of posing a *controlled experiment*

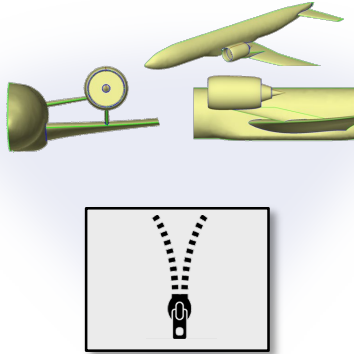


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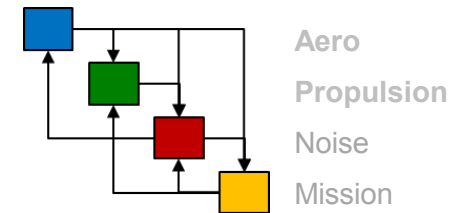


# MDAO Strategy Focus On Efficient Use of Supercomputing Resources

Mathematical techniques for reducing design dimensions



Architecting of multidisciplinary analysis



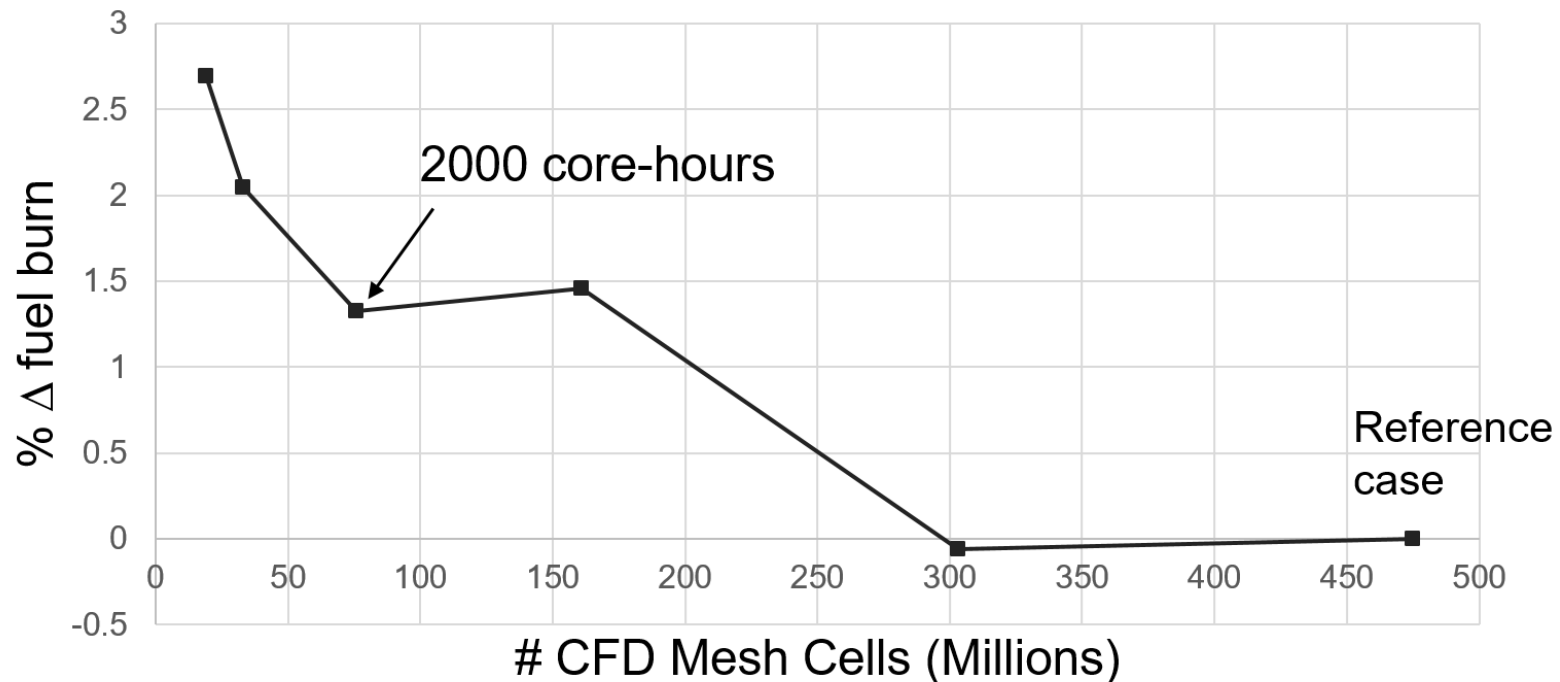
Adaptive sampling methods on supercomputing clusters



[https://en.wikipedia.org/wiki/Pleiades\\_\(supercomputer\)](https://en.wikipedia.org/wiki/Pleiades_(supercomputer))

# CFD Uncertainty Affects Entire Strategy

Ex: Estimated Mission Fuel Burn  
Uncertainty vs CFD Mesh Size

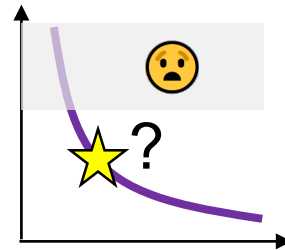


- Published OWN literature uses **0.8 to 30M** cells or grid points
- **Present research requires much higher accuracy and cost**
- (Caveat: grids above are relatively inefficient in favor of robustness over design space)

# Tradeoff: Physics Inaccuracy vs. Unrealized Optimization Benefit

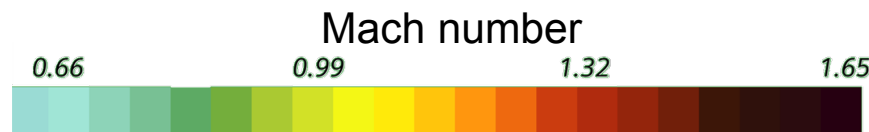
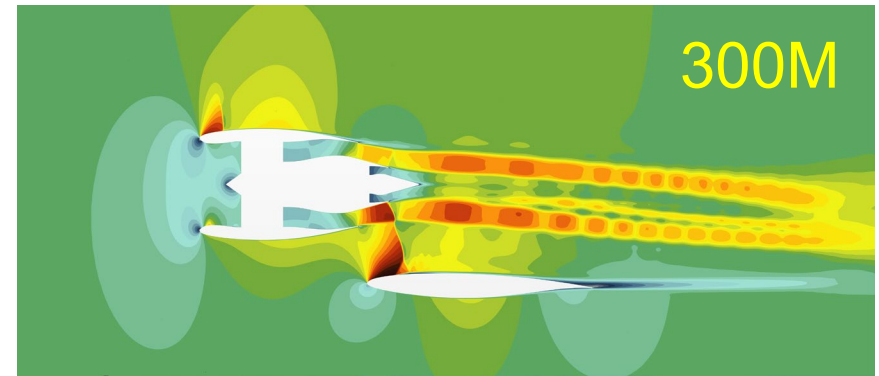
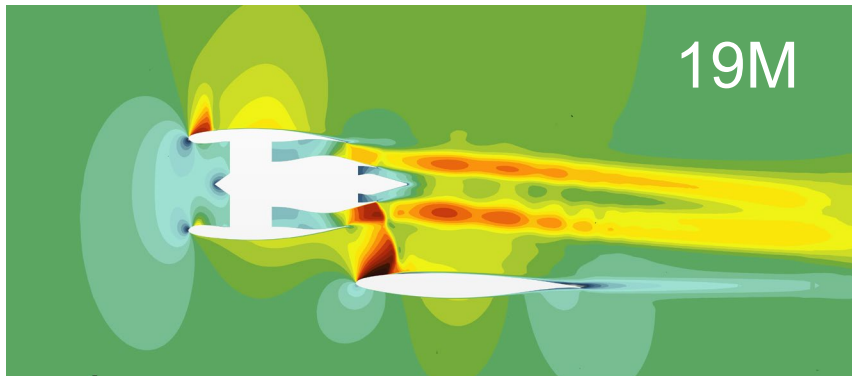
Notional tradeoff  
under constant cost:

CFD error

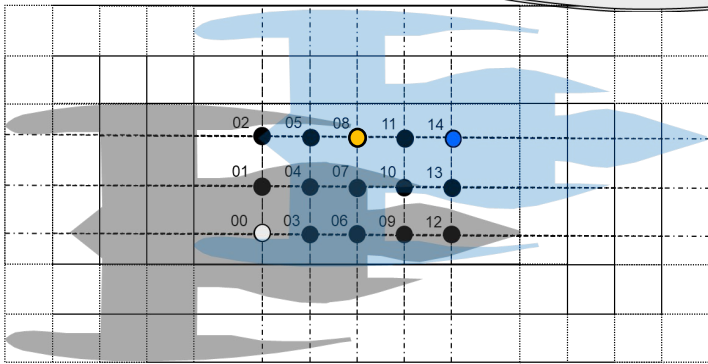
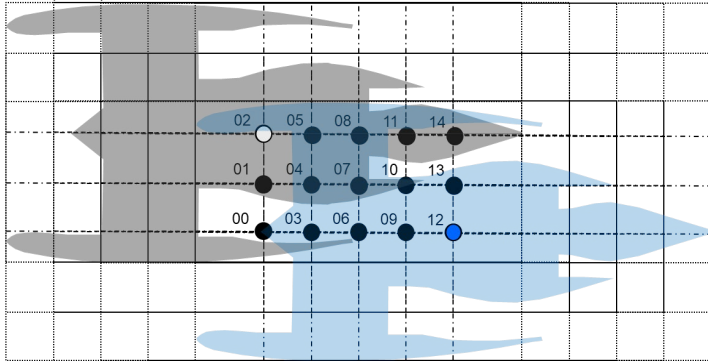


Unrealized OWN  
optimization benefits

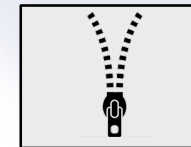
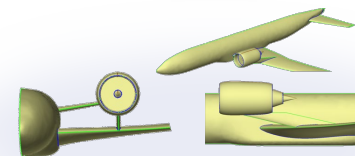
Subtle differences yield  $\Delta$  fuel burn  $> 2\%$



# Recent Progress: First Stage Study Focused on Nacelle Location

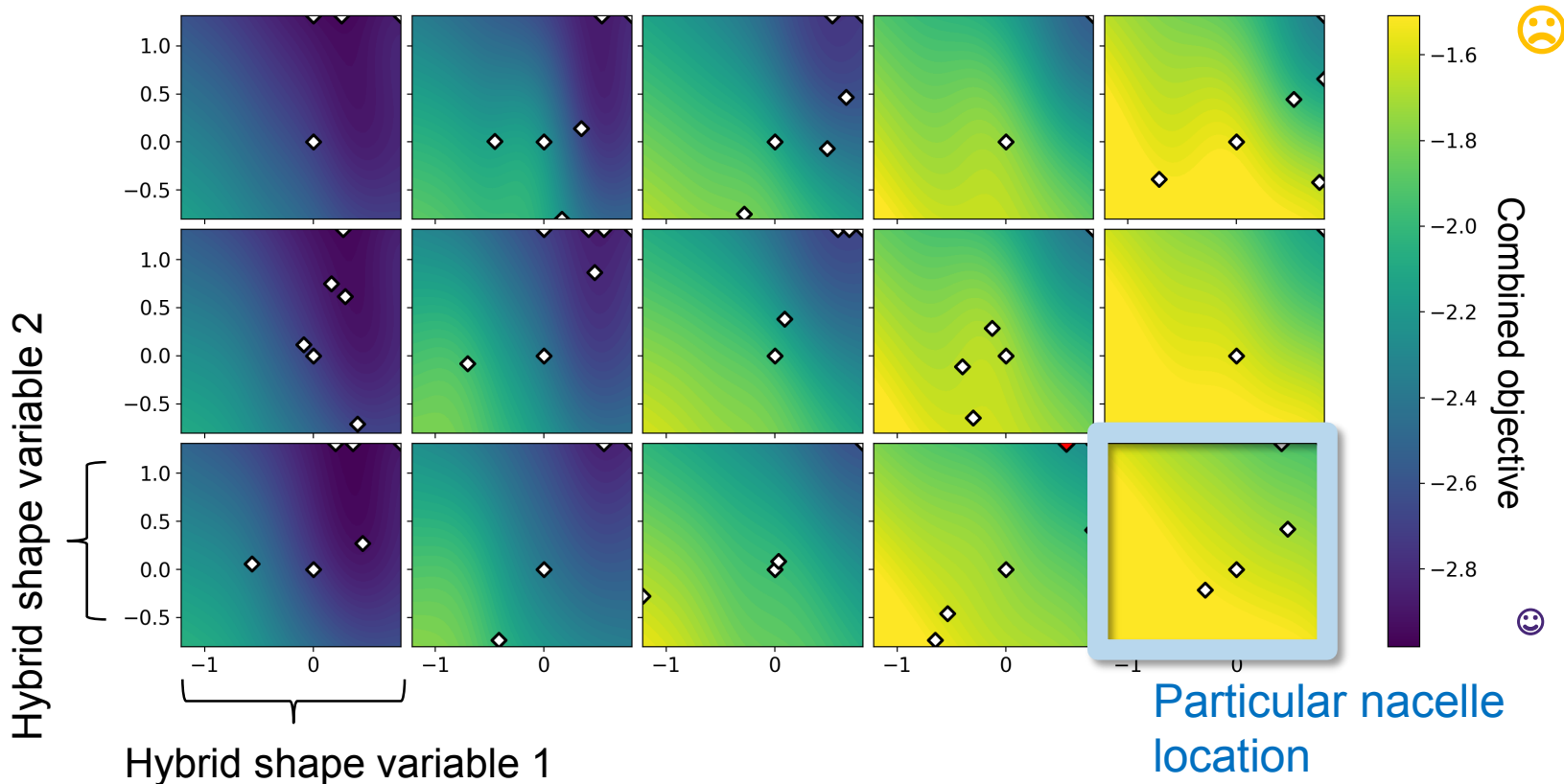
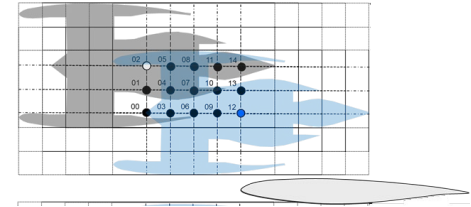


- First stage is to select baseline nacelle location
- For each location:
  - Hybrid (“active surface” wing and nacelle variables) optimized within uncertainty



# Adaptive Sampling Example

- Each color-map “pane” corresponds to an engine location
- Shape variables sequentially sampled within each pane
- Objective function is sum of normalized L/D and -excess thrust





# Next Steps Emphasize on Tight MDAO Integration and Uncertainty Tracking

By June 2022:

- **Larger scale optimization (stage 2)**
  - At fixed nacelle locations, OWN and UWN
  - Tightly coupled CFD and propulsion cycle analysis
- **MDAO integration** with mission analysis
  - Legacy methods incur physics uncertainties that must be tracked!
- **Supporting design sensitivity studies**

